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ABSTRACT

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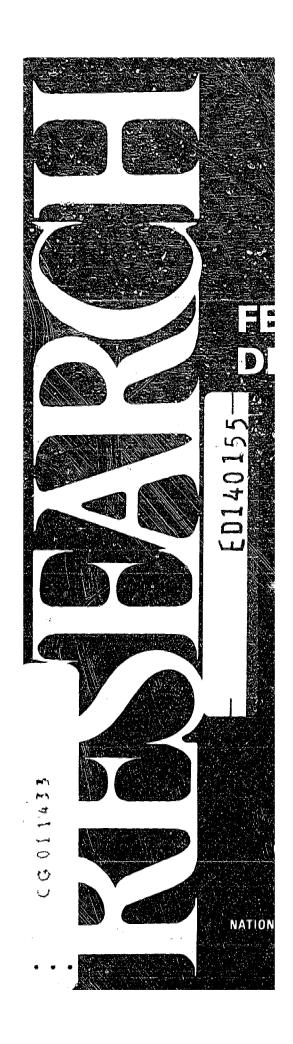
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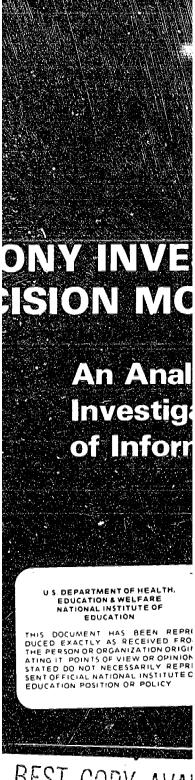
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FELONY INVESTIGATION DECISION MODEL:

An Analysis of Investigative Elements of Information

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FEBRUARY 1977

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ABSTRACT

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Only for robbery was it found feasible to construct a decision model. Primary case-solution factors, e.g., victim knowledge of offender, statistically dominated other, random factors. The findings showed that, unless offender ID was made by responding officers, case solution at the detective level was minimal. Therefore, it was concluded that patrol and investigative functions cannot be viewed as completely separate. Documentation of relevant crime scene information by patrol heavily influences case solution by investigators. The findings reinforced the importance of a national issue: habitual offenders. Analyses of the felony case sample drawn showed 80-88% of the suspects had prior offenses. Confronted by similar experience many police agencies have turned to computer-based M.O.-type investigative systems to assist in tracking and identifying known offenders. However, such systems have yet to demonstrate marked success.



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FOREWORD

For too long, the criminal investigation process has been cloaked in a mystique, and police administrators and researchers have neglected, largely, to address themselves to the development of new models of the investigative components of the service. It has been only during the last few years that we have begun to apply the scientific approach and to inquire concerning technological applications to a process that consumes an inordinate amount of our time and personnel resources.

Rather extensive research has been conducted in the Oakland Police Department in an effort to develop a new investigative model that would serve, among other things, to redefine the methodologies, goals, priorities, and objectives of the criminal investigation process. That research has made it abundantly clear that, to be effective, any new model must be structured around a workable investigative caseload and, to this end, strategies must be developed to identify and minimize the attention given to those offenses that have a low probability of successful clearance.

The work accomplished by the Stanford Research Institute staff during the conduct of the "Felony Investigation Decision Model" study has added significantly to our research efforts. Of great importance to us, the findings suggest that we must reevaluate our traditional thinking concerning the role of the patrol officer in the investigative process, and we must give very careful attention to our training and recording functions to ensure that maximum attention is given to those investigative elements of information that have been shown to be useful in the solution of crimes.

George T. Hart Chief of Police Oakland Police Department

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PREFACE

Over the past decade cynicism has grown with regard to the ability of the police to solve crimes. It is fairly evident that court dockets are crowded, jails are filled, and probation and parole case loads far exceed the ability of corrections personnel to effectively handle the charged, incarcerated, or released felons. All these factors attest to the ability of law enforcement to arrest law violators on a vast scale. But the successive echelons of the criminal justice system have been unable to cope effectively with the intake population.

In undertaking the research reported here we were aware that the police are devoting considerable effort to dealing with repeat offenders; consequently, the research design took into consideration this problem. We were also concerned with the roles of patrol and investigators that influence crime reporting and crime investigation. Although we did not propose to address the causes of the high incidence of crimes in Oakland, the city in which we undertook the research, we recognized the need to maximize the efficiency of investigative resources in handling and solving these crimes. The purpose of the research was to ease the burden of individual investigators who receive a high volume of felony crime reports having a low probability of successful clearance. We deemed it important to find out the actual contribution of computerized data banks to cases that had been cleared.

Our approach was to minimize intuitive judgments on case handling by OPD officers at the patrol and follow-up investigative levels. In other words, we sought to allow massive statistical data "to speak for itself." However, there were many instances that necessitated frequent

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contact with individual investigators and CPD management to interpret our observations and findings. Consequently, we greatly appreciate the support provided by Chief George Hart, Oakland Police Department (OPD), Deputy Chief John Ream, Bureau of Investigation, and Captain John Lothrop, Commander, Criminal Investigation Division (CID). We also extend our sincere appreciation for the time given to answering questions and providing information by many personnel in the CID, Patrol Division, Records and Communications Division, Youth Services Division, and Research and Development Section.

We are grateful to personnel of the Los Angeles, Kansas City (Missouri), Rochester (New York), New York City, and San Diego Police Departments, who provided insight into their respective approaches to the use of computer-based investigation systems and data collection and processing procedures.

We acknowledge the valuable individual contributions to the study made by the following SRI staff and consulting support personnel:

Dr. O. S. Yu, consultant (systems analyst); Dr. P. L. Tuan (senior statistical and computer analyst); J. J. Guidici, consultant (Captain, OPD retired); B. E. Suta (senior operations analyst); J. G. Smyser (policy analyst); and R. Shane and R. N. Schwoegler (data coders, California State University, Hayward, Graduate School of Public Administration).

Finally, the principal investigator would like to honor the memory of Chief John Fabbri, Fremont, California Police Department, who inspired and supported not only this research but prior work undertaken to enhance the criminal investigation function.



SUMMARY

A. Objectives

The objectives of this research project grew out of an SRI study entitled "Enhancement of the Investigative Function." The earl er study developed an insight into the roles of detectives and patrol in conducting burglary investigations. One aspect of the study that appeared to capture the attention of police management nationally was the development of a case follow-up decision model for burglary. It remained to be determined whether the burglary decision model could be usefully applied to car theft and to crimes against persons, where a direct confrontation occurs between victim and offender. Therefore, this project was undertaken to determine the feasibility of structuring case follow-up decision models for certain categories of such crimes. The Oakland Police Department (OPD) consented to be the host agency for the research effort.

In recent years the role of the detective has come under increasing scrutiny. Consequently, in designing this research project, we sought to maximize the efficiency of investigative resources by alternative means. The primary objective of the project was to ease the burden of investigators reviewing a high volume of felony crime reports that have a low probability of successful clearance. The secondary objective of the research was to determine the elements of information leading to offender identification and case solution by investigative personnel, together with the evaluation of computer-assisted investigation systems.



^{*}B. Greenberg et al., "Enhancement of the Investigative Function," Vols. I, III and IV, NTIS PB222-895/896/897, Stanford Research Institute, Menlo Park, California (1972-1973).

B. Overall Findings and Implications

For the felony categories considered in this study--robbery, rape, assault with a deadly weapon (ADW), and car theft, we could realistically construct a decision model only for robbery. Primary case-solution factors--victim knowing the offender in rape and ADW categories, and apprehension of car theft perpetrators being largely effected in an identified stolen car--are so powerful that they statistically dominate other, random elements leading to suspect ID. In fact, our analyses of the four felony categories showed that a large number of cases essentially "solve themselves." By the time a detective receives certain reports, only routine procedures need be followed to apprehend the suspect.

The decision model evolved for robbery reflects the finding that, unless relevant information had been obtained at the crime scene by the responding officer, if the offender had not been apprehended, the chances of the case being solved at the detective level were minimal. The data show that patrol was effecting the larger percentages of case clearances by arrest compared to the CID investigators. Also, except for car theft, such clearances were largely made in less than 8 hours. A conclusion drawn from this observation is that the roles of patrol and investigators cannot be viewed as completely separate and distinct functions. We view patrol as fulfilling not only a crime-suppressant role but also as performing an investigative function. How effectively the patrol officer documents the events of a crime to which he responds will have a definite impact on the case outcome when investigators attempt to pursue the case.

Many facts attest to the ability of law enforcement agencies to arrest law violators on a vast scale. It is evident that court dockets are crowded, jails are filled, and probation and parole case loads are excessive. But these successive echelons of the criminal justice system have been unable to deal effectively with the charged, incarcerated, or

released felons. Thus, this study reinforces the importance of what has now become a national high-priority issue: the finding of a large habitual offender population. In effect, the police are devoting considerable effort to dealing with repeat offenders—who may be more readily identifiable than first-time or transient offenders if police operations are geared to operate on this basis.

Our analyses revealed that, of the persons in our three-month sample of cases drawn who were either last charged or suspected of robbery, 81% had one or more prior offenses; for assault, 80% had prior offenses; for auto theft. 86% had prior offenses; and for rape, 88% had prior offenses.

Confronted with these facts, many police agencies have turned to computer-based investigative systems to assist in tracking and identifying known offenders. It appears, however, that computers have not demonstrated marked success in assisting police in solving modus operandi (M.O.) investigation problems.

On the basis of the information gleaned from the literature and the data generated by our research, we have concluded that the utility of EDP suspect/event-oriented systems is highly dependent on a massive data collection and compilation effort. However, the collection of finely detailed descriptors on personal appearance and events is not only expensive and time-consuming, but may actually be counterproductive.

We question the wisdom of burdening patrol officers with extensive precoded check-off forms, with which several police agencies have been experimenting. We further question whether victims are able to respond adequately to a long list of questions after having been subjected to the trauma of an assault or an armed robbery. The main objective of patrol—to ensure the safety of the victim and quickly ascertain what information can be derived to hasten the offender's apprehension—can

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be thwarted by undue delay in running over a list of data that are likely to be useless.

The results of the research we undertook have, in effect, posed at least three crucial questions that police investigators and planning and funding agencies should consider in their quest for investigative aids:

- What elements of information can police investigators realistically expect to obtain regarding a crime event and the personal characteristics of the offender?
- What are the best procedures for establishing and preserving a logically structured data base that can recall the information that will materially aid the investigator in solving a given crime?
- Is it realistic to expect that the classical concept of M.O. can be developed for automated data processing systems to enable recognition of a distinctive crime commission pattern exhibited by a given offender?

We suggest that these questions and other investigative issues raised from the research findings, e.g., the relationship of patrol and investigators, can best be addressed in the context of a workshop involving LEAA and leading law enforcement agencies concerned with the interrelationship between investigative and patrol operations and with the contribution that computer technology can make toward controlling the criminal population.

Too frequently, a research report gathers dust on a recipient's bookshelf. But by considering important, or at least controversial, findings in a workshop, participating agencies might find more reason to become part of the creative policy and decision-making processes that can impact on the nation's growing crime rate.

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Investigation Decision Models: Robbery and Burglary

The analytic methodology used for constructing an investigation decision model for robbery was followed for each of the other felony categories. However, the nature of the other three felonies, and the manner by which such cases are solved by the OPD, precluded the development of additional decision models. We describe in detail the analytic process for robbery leading to the development of the model. Observations on case handling and conclusions short of a decision model, however, are summarized for all the categories.

Computer subprograms were used that systematically narrowed the large number of variables analyzed to those showing a significant level of occurrence in felony case clearances. The professional detective might construe this narrowing process as eliminating from consideration pieces of information that might prove to be valuable in solving a case. While we concede that this is a distinct possibility for random cases, there is a larger issue that needs to be addressed concerning the general procedures that are effective in handling the high volume of crimes. This issue centers on the types of information categories, e.g., facial features, and the numbers of "permanent" and variable time-sensitive descriptors that may be critical to identify a suspect and that should be captured in a preliminary report of investigation.

Since we recognized that the same kinds of information would appear in both the cleared and the uncleared cases, the statistical technique used was to cluster the various data elements contained in both types of cases and to weight them in accordance with their degree of association with the cleared cases.

The four subcategories of robbery considered were combined (armed, strong-arm, theft from person, and purse snatch) so as to construct the robbery investigation decision model by using linear discriminant analysis.

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This procedure strengthened the discrimination power of the data category elements to enable the construction of a model with a high predictive probability that a case taken at random could be correctly classified as cleared or uncleared. The analysis produced a numerical value for each piece of information contained in a case report. This value shows the relative contribution of that piece of information to case clearance as compared to all the other pieces of information.

The reader must keep in mind that the decision rule, shown in Table S-1, is based on the OPD's operational practices affecting case handling (e.g., whether the reporting officers recorded all useful information) and consequently affecting the manner by which cases are cleared. Other police departments may not have similar policies, procedures, and capabilities. Usage of the decision model must thus be carefully considered in light of a specific agency's operational procedures. It can be seen that our decision model contains a number of items of information that result from the preliminary case enrichment procedures routinely performed by the Crime Analysis Section (CAS) of the OPD Criminal Investigation Division (CID). This implies that the case disposition screening process should take place at some time after certain basic investigative procedures, e.g., license number checks, have been pursued. Thus, the important consideration is that this model should be considered in a dynamic mode; i.e., the weighted elements should be checked throughout the investigating phase of the case. Should a suspect then not be identified, the case can be realistically set aside as being unsolvable.

Of the cases in the sample, 90% were correctly classified as cleared or uncleared by the classification function derived from the discriminant analysis. This is reflected by the relative scaling in the decision model. Further analysis indicated that the 10% misclassification was not as serious as might first appear. Most of the cleared cases that the

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Table S-1

ROBBERY INVESTIGATION DECISION MODEL

Information Element	Weighting Factor
Suspect named	10*
Suspect known	10*
Suspect previously seen	10*
Evidence technician used	10
Places suspect frequented named	10*
Physical evidence	
Each item matched	6.1
Vehicle registration	
Query information available	1.5
Vehicle stolen	3.0
Useful information returned	4.5
Vehicle registered to suspect	6.0
Offender movement description	:
On foot	0
Vehicle (not car)	0.6
Car	1.2
Car color given	1.8
Car description given	2.4
Car license given	3.0
Weapon used	1.6

INSTRUCTIONS

- Circle the weighting factor for each information element that is present in the incident report.
- (2) Total the circled factors.
- (3) If the sum is less than 10, suspend the case; otherwise, follow up the case.
- (4) Weighting factors do not accumulate; i.e., if both the auto license and color are given, the total is 3.0 not 4.8.

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^{*} These values as calculated actually exceed the threshold of 10. The values provided here are conceptually simpler and make no difference in the classification of groups.

decision model had predicted would remain uncleared were cleared without OPD investigation. In only two cases, investigation by OPD detectives ultimately led to the identification of a suspect despite the small amount of information initially available in the reports.

For comparison, Table 3-2 shows the burglary case disposition decision rule developed in a prior SRI research project. Attention is

Table S-2
BURGLARY CASE DISPOSITION DECISION RULE

Information Element	Weighting Factor
Estimated range of time of	
occurrence	
Less than 1 hour	5
1 to 12 hours	1
12 to 24 hours	0.3
More than 24 hours	0
Witness's report of offense	7
On-view report of offense	1
Usable fingerprints	7
Suspect information developed	
description or name	9
Vehicle description	0.1
Other	0

INSTRUCTIONS

Total score

- Circle the weighting factor for each information element that is present in the incident report.
- (2) Total the circled factors.
- (3) If the sum is less than or equal to 10, suspend the case; otherwise, follow up the case.

From: B. Greenberg et al., op. cit., Vols. I and IV.



drawn to the similarity of the variables in the two rules and their relative weights in contributing to case clearances. For both models, the witness viewing or victim involved in the crime provides the most useful information leading to case clearance. But for the robbery cases, apart from the naming of the suspect, which is also a dominant element in burglary case clearance, vehicle information is the next-most-important information element leading to suspect ID.

The predictive accuracy of the burglary model was found to vary widely among the agencies whose cases were analyzed. The explanation for the wide variation is simply that the various agencies involved had inconsistent policies governing the criteria by which a burglary case was cleared. Consequently, the robbery model developed on the basis of the OPD's policies must be carefully considered by other agencies who may desire to apply it. Clearance criteria will affect an agency's effective use of the model as a screening tool.

D. Offender Characteristics

We tracked the criminal histories of suspects identified in the sample of felony crimes analyzed. It is clear that the OPD is processing a large recidivist criminal population. Table S-3 illustrates past charged criminal offenses classified into 17 categories. This table was developed to examine the hypothesis that repeat offenders in the four felony categories would show different patterns of past offenses.

Persons whose most recent offenses were ADWs had high past incidence of burglary, theft, other assault, narcotics and dangerous drugs, vehicle law violations, and other offenses.

Table S-3
OFFENDERS' PRIOR CRIMINAL HISTORIES

Prior Offense	Мо	st Recent O	ffens e Ca tego	ry
Classification	ADW	Robbery	Car Theft	Rape
No prior	19.8%	18.7%	14.2%	12.5%
Strong-arm robbery	12.1	14.9	12.4	12.5
Armed robbery	2.8	10.5	5.3	12.5
Felony assault	21.4	13.4	14.8	18.8
Burglary	28.6	46.3	47.3	56.3
Auto theft	14.3	22.4	40.8	25.0
Homicide, willful	2.8	1.5	1.8	0.0
Forcible rape	2'.2	2.2	4.7	12.5
Attempted rape	0.6	2.2	0.0	6.3
Theft, person	0.6	3.7	1.2	6.3
Theft, purse snatch	1.7	3.0	4.1	6.3
Theft, shoplifting	11.0	9.7	21.3	12.5
Theft, other	28.6	38.1	47.9	31.3
Narcotics and drugs	22.5	29.9	29.6	43.8
Stolen property	7.1	9.0	21.3	12.5
Vehicle laws violation	32.4	23.1	32.0	43.8
Other	64.3	58.2	70.4	75.0
Other, not indicated	2.8	11.2	0.6	0.0

Each category shows the percentage of offenders who had previously been charged with each of the 17 offenses.

Persons whose most recent offense was robbery showed high past incidence of burglary, car theft, theft other, narcotics and dangerous drugs, vehicle law violations, and other crimes.

Persons whose most recent past offense was car theft had the highest percentage of past car theft, shoplifting, theft other, and possession of stolen property. They also had high past incidence levels of burglary, narcotics and dangerous drugs, vehicle law violations, and other crimes.

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Persons whose most recent past offense was rape had the highest percentage of past burglary, rape, narcotics and dangerous drugs, vehicle law violation, and other crimes. They also showed a high past incidence of car theft.

Other summary data revealed the following:

- · Repeat offenders averaged more than 7 prior offenses.
- Over 80% of the offenders were black. (The population of Oakland is approximately 43% black.)
- · Over 90% of the offenders were male.
- On the average, the repeat offenders in the four categories had criminal records of 7.4, 8.3, 10.8 and 12.1 years.
 These figures are associated with car theft, robbery, ADW, and rape, respectively.
- The persons whose most recent offense was car theft had had the highest average number of offenses per year. The average was 1.8 offenses charged per year, contrasted to the ADW and robbery offenders who averaged 1.1 and 1.3 offenses charged per year, respectively.

The tabulation below shows the number of offenders whom we analyzed in the three-month sample in the four felony categories and the total number of offenses charged for this offender population.

	Most	Recent Of	<u>fense Ch</u>	arged
	Robbery	ADW	Rape	Car Theft
Number of offenders	134	183	16	169
Total offenses charged	836	1,067	129	1,269

It is quite evident that the offender population in our sample had committed a significant number of multiple offenses.

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E. Conclusions on the Implications of Uniform Descriptors for Investigative Application

The concept of the computer manipulating vast amounts of data and spewing out all sorts of information has captured the imagination of hardpressed law enforcement agencies as an aid in tracking and identifying felony crime offenders. A 1972 report published by the International City Management Association (ICMA)* predicted that computer procurement for criminal investigation applications will more than quadruple over the next few years. The ICMA further reported that "the surface has only been scratched when it comes to the use of the computer for criminal investigation." While the ICMA referred to agencies experimenting with computerized M.O. systems, they acknowledged that the law enforcement community is divided in their views on the utility of such systems. The article also referred to the assignment of cases to investigative officers on the basis of the probability of cases being solved.

Felony crime solution factors illuminated in this study show that only a small number of investigative elements of information have proved generally useful in crime solution. This finding may incur anathema from several notable police agencies that have gone to great lengths to attempt to capture vast amounts of personal appearance and M.O. information in anticipation of increasing the likelihood of offender identification and apprehension.

Our findings on criminal activity patterns reinforce the ICMA statements of the split views of the law enforcement community. Offenders do not tend to display consistency. They engage in a multitude of crimes, and consequently law enforcement must deal with repeat offenders across a broad spectrum of crimes. This fact alone should encourage police



ICMA, "Use of Computers by Police: Patterns of Success and Failure,"
International City Management Association, Washington, D.C. (1974).

departments to develop coordinated efforts to cross over investigation specialization areas and to attempt to construct well-thought-out offender identifiers.

Our recent experience in attempting to assess the utility of a computer-based known-offender investigation system in the host agency raised a critical question: Why did the system not produce better results than those we were able to discern? For example, out of the 205 cases we processed that the CAS requested be run, only 8 useful leads for suspect ID were found. The vehicle subfile system, whose data base input probably limited its utility, produced only 2 useful leads out of 28 runs requested.

The answer to this critical question probably lies in three areas. The original data drawn from a known-offender arrest record file are old; consequently the descriptors derived from more current incident reports may be incompatible with the data base, resulting in a large error due to mismatch. A second problem area may be the operator, who inadvertently causes suppression of possible hits by omitting certain data or not allowing for a sufficiently wider range of, say, possible hair color or hair length. A third problem area, probably a major technical failing, may lie in the software program, which may not have been accurately designed initially. (We also learned that a physical break in the optical lens scanner caused the random search process to produce large errors in hits.)

Although many agencies can cite random successes in developing suspect ID by means of suspect/event-oriented computer based systems, given the present state of the art, and human judgment considerations, we conclude that collection of unlimited numbers of information elements for computer processing is not a panacea for crime solution.



In approaching the attainment of our primary goal, we recognized that a major by-product of the research effort would be the identification of descriptors of events and offenders useful in case solution. We observed that there is an extremely wide variation of format in police incident report forms, not only within a county in California, but statewide and nationally. This variation reveals basic differences in comprehension of the types of information that are crucial for crime reporting, investigation, and prosecution purposes.

The police incident forms vary considerably in complexity. General agreement in critical descriptors is a necessary prerequisite if the best offerings of computer technology are to be effectively utilized.



CHAPTER I. INTRODUCTION

A. Decision Model Conceptual Background

This research project grew out of a study undertaken by SRI in several police agencies in Alameda County, California.* That study examined the roles of detectives and patrol in conducting burglary investigations. The major objectives were to develop a checklist of activities and then a handbook, primarily for the guidance of patrol officers in gathering the most useful information leading to the identification and arrest of an offender and successful case closure. One aspect of this study that appeared to capture the attention of police management nationally was the development of a case follow-up decision model. This case selection model, in essence, is a set of weighted variables or elements of information that, if present in a burglary report at a predetermined numerical level, will enable the case outcome to be predicted with a high degree of certainty.

The model had been validated in the original agencies participating in the experimental program. It was also validated independently by an Oakland Police Department (OPD) consultant team using OPD burglary reports. The surprising result was that 90% accuracy was reported as to whether random cases could be solved—and therefore should be followed up—or could not be solved—and consequently should be set aside, so as to min—imize the paperwork burden on investigators.



B. Greenberg et al., "Enhancement of the Investigative Function," Vols. I, III and IV, NTIS PB222-895/896/897, Stanford Research Institute, Menlo Park, California (1972-1973).

While the independent OPD validation of the model was encouraging, we had found a wide disparity of confidence in the case clearance probability predictions among the smaller agencies from which our original burglary case samples had been drawn. The accuracy of the prediction levels was clearly associated with varying standards for arrest and case clearance in the participating agencies. The analyses conducted showed clearly that certain agencies stressed some aspects of investigative practices that others did not--and could not, owing to budgetary constraints or policy considerations.

Because the question remained whether the burglary case selection model could be useful for application to crimes against persons, where a direct confrontation occurs between victim and offender, this project was undertaken to analyze crimes against persons and car thefts. The OPD consented to be the host agency for this research.

B. Crime, Socioeconomic, and Demographic Characteristics of Oakland

The 1974 Preliminary Annual Release of the Uniform Crime Reports showed that, although Oakland has a historically high crime rate, the overall crime index for Oakland decreased 3% between 1973 and 1974, whereas for cities having populations of 250,000-500,000 (the group into which Oakland falls), the crime index showed a general 13% increase. The overall national crime index rose by 17%. Except for aggravated assault, Oakland has been going against the national trend for the seven major felony crimes. Table I-1 shows the comparison.

Oakland has a population of approximately 350,000 and has experienced a slight decline in total population since the 1970 census. The city is changing ethnically and is characterized by emigration of whites and immigration of blacks. The educational level of the citizens of Oakland has shown continual improvement. Females are slightly in the majority and are increasingly entering the labor force. Male employment,

Table I-1

CRIME INDEX COMPARISON--OAKLAND AND OTHER CITIES WITH POPULATIONS BETWEEN 250,000 AND 500,000

4	<u>Year</u>	Crime Index	Murder	Forcible Rape	Robbery Oakla	Aggravated Assault	Burglary	Larceny Theft	Motor Vehicle Theft
	1973	41,595	100	220	2,879	1,853	14,734	.17,063	4,746
	1974	40,507	78	246	2,883	2,175	14,144	16,702	4,279
	Percent Change	- 3%	-22%	+1.2%	+0.1%	+17%	=4%	- 2%	=10%
1	Other Cities								
	Percent Change	+13%	+1%	+14%	+13%	+5%	+16%	+16%	= 27/a

on the other hand, has been declining. This reduction in male occupation may account for the high unemployment rate as the city gradually shifts to predominantly white-collar jobs.

The social fabric of Oakland appears to be undergoing a fundamental transition. The city's residents are primarily young (between the ages of 18 and 34) and single, and family units tend to be smaller than formerly. The residents are tending to leave single-family housing units and to move to apartments. The changing socioeconomic picture is not appreciably different from that of most other comparable urban areas. (See Appendix A for a more complete discussion.)

C. Oakland Police Department and Criminal Investigation

To cope with the high level of reported crimes, the OPD introduced many important innovations to improve its delivery of services, particularly use of computer-aided systems and a recent restructuring of patrol operations. (See Appendix B for a detailed description of the OPD and the reporting forms analyzed in this study).

The Criminal Investigation Division (CID) is the division of the OPD of direct concern to this project. Although certain functions and procedures have been introduced to enhance the overall efficiency of CID investigational operations, the basic personnel staffing structure and the responsibilities of the detective force do not appear to have been appreciably altered in recent years. The OPD has introduced a Crime Analysis Section (CAS) into the CID and provided a staff of trained civilian computer operators to process crime reports for "enrichment" by interrogating various data banks in the OPD, Alameda County, the California Department of Justice, and the FBI National Crime Information Center (NCIC). The CAS role clearly reflects the recognition that routine data file search functions need not be delegated solely to a skilled detective.



In recent years the role of the detective has come under increasing scrutiny, particularly since the overall national crime rate has soared despite the large amounts of funding provided to law enforcement to bring the crime rate under control. While we have not addressed the probable causes of the rise in crimes, we do recognize the need to maximize the efficiency of investigative resources by alternative means. The primary objective of this research project was to ease the burden of investigators who review a high volume of felony crime reports having a low probability of successful clearance. Consequently, we undertook to analyze a stratified sample of cleared and uncleared cases for four felony categories: robbery--armed, strong-arm, theft from person, and purse snatch, rape-attempted and forcible, assault with a deadly weapon (ADW), and car theft. Our purpose was to determine the feasibility of structuring case follow-up decision rules on the basis of our prior experience in constructing a burglary decision model.

Investigators already apply subjective judgment in determining which cases look sufficiently promising to pursue. But there are basic inefficiencies in relying solely on individual experience and judgment to select the cases to be pursued. The task of reviewing reports for such high-volume crimes as burglary, robbery, assault, and car theft is tedious. The large case-load backlogs piled onto investigators are distracting. Moreover, it has become increasingly clear that, of the felony types analyzed, the majority of the cases cleared have been solved by patrol.

The paperwork generated by patrol (on all cases, whether cleared at the scene or not) shows a tremendous variation in the quality of the information of record that is transmitted ultimately to a detective for possible follow-up. It should be noted that the OPD, in contrast to other, smaller departments in Alameda County, minimizes the involvement of patrol in crime scene investigations. Consequently, the information



secured from the immediate crime scene, for most crime categories, is limited to that which the responding patrol officer is able to secure quickly from the victim or witnesses. It became evident that, unless relevant information is obtained (by patrol or an evidence technician) that will enable further leads to be pursued when the initial report has been filed and passed to the CID, the chances of the case being solved are minimal.

This observation lead to another objective of the study: determination of the elements of information that facilitate identification and apprehension of the offender at the scene, or fleeing from it, and the elements of information that contribute to case solution by investigative personnel. Basically, the initial problem begins at the crime scene, with the ability of the first officers or evidence technicians arriving there to secure relevant investigative information. How efficiently this task is accomplished largely determines the case outcome.

Our approach was to minimize anecdotal examples and intuitive judgment on the case handling by police investigators—at both the patrol and the follow-up investigation levels—by analyzing on a statistical basis the factors that have significantly contributed to case clearance. To do so, we used an extensive, statistically based analytic methodology with the intent to: determine primary elements of information that would enable construction of a case follow-up decision model; emphasize the information elements that trained patrol officers can realistically be expected to secure, assuming cooperative and observant victims and witnesses; and identify the investigative processes that appear to materially assist investigators in identifying offenders.

D. Summary of Felony Crime Statistical Analyses

Each of the four major felony categories analyzed is discussed in a separate chapter. Following is a summary of the levels of occurrence



of each of the felonies during the three-month sample period, as well as an explanation of how to read the cross-tabulation tables.

1. Robbery.

a. Armed robbery. During July, August, and September 1974 330 armed robberies occurred (see Table I-2). Of these 42 cases (12.7%) were cleared, 9 (2.7%) were cleared-other, and 279 (84.5%) were uncleared. The matrices are interpreted as follows. The column headings indicate the type of case disposition officially made by the CID investigators. The rows indicate the clearance category. The first matrix cell shows that 23 armed robberies were cleared by arrest and prosecution Other entries show the various clearance categories. The Row Total column shows that 42 cases were cleared out of a total of 330 robbery cases occurring in the same period, for a cleared rate of 12.7%.

The second line in the cleared matrix cell under the Arrest and Presecution column shows that 54.8% of the cleared cases were classified by this category.

The third line of the Arrest and Prosecution column indicates that all (100%) of this column of cases were classified in this manner (this is a column percentage value).

The fourth line indicates that 7% of all armed robbery cases were cleared under the Arrest and Prosecution classification.

Cleared cases are those for which the OPD took any formal disposition other than "Complainant Refuses To Prosecute" or "Complaint Refused by District Attorney." Cleared-other cases are those for which the OPD took a "Complaint Refused by District Attorney" or "Complainant Refuses To Prosecute" disposition and a suspect was named. Uncleared cases are those with a formal disposition of "Complainant Refuses To Prosecute" where a suspect was not named, cases where an investigator filed the case without a disposition, and cases where there was no evidence of investigative attention. Appendix D gives a complete discussion of this breakdown.

Table I-2

ROBBERY, ARMED: CLEARANCE BY CASE DISPOSITION

ı		Arrest	Complainant	Prosecuted	Compl ^u int	Prosecuted	Turned Over t ^o	Notice		
	No	and	Refuses To	tor Another	Refu ^{#6} d	ph Onto	Juvenl Je	ra Aliter		
<u>Status</u>	<u>Disposition</u>	Prosecution	Prosecute	Offense	by Dy.A:	Jane Denie	Authority	to to	<u>Total</u>	
Cleared						And the second				
Count	Ō	23	: ()	2	Q	1	9	1	42	
% of row	0%	54.8%	0%	4.8%	0%	1/4	21.4	- #/		
% of column	0	100,0	0	100.0	O	16.		2,4%		
% of total	0	7.0	0	0.6	********* 0	100.	100.0 2.7	0,00.0 0,3	12.7%	
Cleared-Other								âi.		
Count	Ō	· Q	8	0	1	0	Λ		ğ	1 - 12 s 12 s
% of row	0%	0%	88.9%	Ô [¶] /2	,, 1//	0%	0 0*	0	•	
% of column	0	0	28.6	0	11,1"	0	-	0%		
% of total	0	0	2,4	0	0.3	Õ	0 0	0 C	2,7%	
Uncleared					i i i			•		
Count	259	0	20	0	0	0	0	Я	279	
% of row	92.8 %	0%	7.2%	0%	0%	ø%	0%	0 0%		
% of column	100.0	0	71,4	0	. 0	0	Ūr Q	•		
% of total	<u> 78.5</u>	0	6.1		_9	ġ	_0	\ ()	84.5%	
Total Count	259	23	28	2	1	1	9	,	330	
% of Cases	78.5%	7.0%	8.5%	0.6%	0.3%	2.1%	2.7%	ļ 6 19	100.0%	á í
		į		¥.	ν,	1.	4,1°	0.3%	•	3, (

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The Cleared-Other matrix cell is read in the same manner. Note that under the column heading Complainant Refuses To Prosecute [Category 03 in Figure B-9(a) in Appendix B] we have listed the cases where a suspect was named--8 cases cleared-other, for an overall percentage of 88.9% for all cases cited as cleared-other. Line 3 in this column shows that 28.6% of the Complainant Refuses To Prosecute cases fell into the cleared-other category. Line 4 in this column shows that 2.4% of all armed robbery cases were classified as cleared-other--Complainant Refuses To Prosecute, and suspect had been named.

The Uncleared matrix cell has two major column totals. The first column, titled No Disposition shows the cases and their percentages that remained uncleared. Under the Complainant Refuses to Prosecute column, we classified all such OPD cases, whenever no suspect had been named, as uncleared.

All cross tabulations subsequently presented in this report can be interpreted as explained above.

- b. <u>Strong-arm robbery</u>. Of the 275 cases of strong-arm robbery sampled (see Table 1-3), 36 (13.1%) were cleared; 11 (4%) were cleared-other; and 228 (82.9%) remained uncleared.
- c. Theft from person. Of the 110 thefts from person for the time period (see Table I-4), 13 (11.8%) were cleared; 3 (2.7%) were cleared-other; and 94 (85.5%) were uncleared.
- d. <u>Purse snatch</u>. Of the 103 purse snatches for the time period, 10 (9.7%) were cleared, and 93 (90.3%) were uncleared (see Table I-5). No purse snatches fell into the cleared-other category.



Table I-3

ROBBERY, STRONG-ARM: CLEARANCE BY CASE DISPOSITION

						i			ار با		
		No	Arrest And	Complainant Refuses To	Prosecuted for Another	Complaint Refused	Prosecuted by Outside	Reprimanded and	Turned Over to Juvenile	Notice to	
: .	Status	Disposition	Prosecution	Prosecute	Offense	by D.A.	Department	Released	Authority	<u>Appear</u>	<u>Total</u>
	Cleared			:					1.1	Á	i nt
	Count	0	20	0	ļ	Ď	1	<u> </u>	11	2	36
•	% of row	0%	55.6%	0%	2.8%	0%	2.8%	2.8%	30.6%	5.6%	
	% of column	0	100,0	0	100.0	0	100.0	100.0	100.0	100.0	1 A 1 B
	% of total	Ō	7.3	0	0.4	0	0,4	0.4	4.0	0.7	13.1%
. ;		•									
	Cleared-Other	ī								_	
	Count	Ō	0	9 :	0	2	0	0	, 0	0	11
	% of row	0%	0%	81.8%	0%	18.2%	0%	0%	0%	0%	
	% of column	0	0	60.0	Q	100.0	0	0	0	0	: ***
	% of total	0	0	3.3	Ō	0.7	0	0	0	0	4.0%
	To start the second				÷						
	Uncleared										200
	Count	22 <u>2</u>	Ō	6	0	0	Ō	0	0	0	228
	% of row	97.4%	0%	2.6%	0%	0%	0%	0%	0%	. 0%	
	% of column	100.0	0	40.0	0	Ď	0	0	0	0	
	% of total	<u>80.7</u>	0	2.2	0	<u> </u>	<u> </u>	0	<u> </u>	0	82.9%
											
	Total Count	222	20	15	į	2	1	Ţ	11	2	275
	% of Cases	80.7%	7.3%	5.5%	0.4%	0.7%	0.4%	0.4%	4.0%	0.7%	≈ 100 .0 %

Table I-4

ROBBERY, THEFT FROM PERSON: CLEARANCE BY CASE DISPOSITION

Status	No Disposition	Arrest and Prosecution	Complainant Refuses To Prosecute	Turned Over to Juvenile Authority	Notice to Appear	Total
Cleared			,			•
Count	0	8	0	4	1	13
% of row	0%	61.5%	0%	30.8%	7.7%	
% of column	0	100.0	0	100.0	100.0	r
% of total	О	7.3	0	3.6	0.9	11.8%
Cleared-Other						
Count	0	О	3	0	О	3
% of row	0%	0%	100.0%	0%	0%	
% of column	0	0	33.3	0	0	
% of total	o	. 0	2.7	О	0	2.7%
Uncleared						
Count	88	0	6	Ō	0	94
% of row	93.6%	0%	6.4	0%	0%	
% of column	100.0	0	66.7	0	0	
% of total	80.0	0	5.5	0	0	85.5%
Total Count	88	8	.9	4	1	110
% of Cases	80.0%	7.3%	8.2%	3.6%	0.9%	100.0%

Table I-5

ROBBERY, PURSE SNATCH: CLEARANCE BY CASE DISPOSITION

Status	No <u>Disposition</u>	Arrest and <u>Prosecution</u>	Complainant Refuses To Prosecute	Turned Over to Juvenile Authority	<u>Total</u>
Cleared					
Count	0	6	0	4	10
% of row	0%	60.0%	0%	40.0%	
% of column	0	100.0	O	100.0	
% of total	0	5.8	0	3.9	9.7%
Uncleared					
Count	92	0	1	0	93
% of row	98.9%	0%	1.1%	0%	
% of column	100.0	Q	100.0	, 0	
% of total	<u>89.3</u>	0		0	90.3%
Total Count	92	6	1	4	103
% of Cases	89.3%	5.8%	1.0%	3.9%	100.0%



- 2. Rape. Only 65 cases of rape were reported during the period. Of these, 16 (24.6%) were cleared; 12 (18.5%) were cleared-other; and 37 (56.9%) were uncleared (see Table I-6). Ten of the uncleared cases had been cleared as Complainant Refuses To Prosecute. This constitutes 29.7% of the cases in the uncleared category.
- 3. Assault with a deadly weapon. The 413 ADWs for the time period (see Table I-7) were cleared at a much higher rate than any of the other crimes investigated. Of these ADWs, 206 (49.9%) were cleared; 147 (35.6%) were cleared-other; and only 60 (14.5%) fell into the uncleared category.
- 4. <u>Car theft</u>. Car theft was the highest-volume crime coded and also had the lowest clearance rate (see Table I-8). Of 1187 car thefts for the time period, 104 (8.8%) were cleared; 38 (3.2%) were clearedother; and 1045 (88.0%) were uncleared.



Table I-6

RAPE: CLEARANCE BY CASE DISPOSITION

and the second s	<u>Status</u>	No <u>Disposition</u>	Arrest and <u>Prosecution</u>	Complainant Refuses To Prosecute	Complaint Refused by D.A.	Turned Over to Juvenile Authority	D.A. <u>Citation</u>	Notice to <u>Appear</u>	Total
	Cleared				5	, 1	i	1	 16
	Count	0	8	0	()	. ↓ A= EN	£ 79/	6.3%	**
٠	% of row	0%	50.0%	0%	0%	37.5%	6.3%		
	% of column	0	100.0	0	a	100.0	100.0	100.0	24.6%
	% of total	0	12.3	, ()	Ō	9.2	1.5	1.5	44 • W /g
	Cleared-other	t			_	ō	٨	0	12
	Count	Ó	0	10	2	Ō	0	•	14
الرا	% of row	0%	0%	83.3%	16.7%	0%	0%	0%	
	% of column	Ď	Ô	47.6	100.0	0	0	0	ia cò
	% of total	0	0	15.4	3.1	ĵ	0		18.5%
ŕ	Uncleared		1.3					_	
	Count	26	0	11	0	0	0	0	37
	% of row	70.3%	0%	29.7%	0%	0%	0%	0%	
	% of column	100.0	Ó	52,4	0	0	0	Ō	
£ *	% of total	40.0	0	<u>16.9</u>	0	0	0	<u></u> ()	56.9%
	Total Count	26	8	21	2	6	1	1	65
	% of Cases	40.0%	12.3%	32.3%	3,1%	9.2%	1.5%	1.5%	100.0%

Table I=7

ADW: CLEARANCE BY CASE DISPOSITION

Status	No <u>Disposition</u>	Arrest and <u>Prosecution</u>	Refuses 'to	Complaint Refused by D.A.	Reprimanded and Released	Turned Over to Juvenile Authority	D,A. <u>Citation</u>	Notice to <u>Appear</u>	Total
Cleared		ŧ		:					
Count	Ō	158	0 '	0	3	20	15	10	206
% of row	0%		· /- 0%	0%	1,5%	9.7%	7.3%	4.9%	
% of column	0	100.0	0	0	100.0	100.0	100.0	100.0	
% of total	0	38.3	0 .	0	0.7	0.7	3.6	2.4	49.9%
Cleared-other		·							
Count	0	٠٠, ٥	138	9	0 .	0	. 0	0	147
% of row	0%	0%	93.9%	6.1%	0%	0%	0%	0%	
% of column	. 0	0	91.4	100.0	0	0.	0	0	
% of total	Ō	. 0	33.4	2.2	0	0	0	0	35.6%
Uncleared									
Count	47	Ō	13	Ô	Ŏ	Q	0	Ō	60
% of row	78.3%	0%	21.7%	0%	0%	0%	0%	0%	
% of column	100.0	Ō	8.6	Ó	Ò.	0	0	0	
% of total	11.4	<u> </u>	3.1	0	0	0	0	0	14.5%
Total Count	47	158	151	9	3	. 20	15	10	413
% of Cases	11.4%	38.3%	36.6%	2.2%	0.7%	4.8%	3.6%	2,4%	100.0%

Table I-8

CAR THEFT: CLEARANCE BY CASE DISPOSITION

Status	No Disposition	Arrest and <u>Prosecution</u>	Complainant Refuses To Prosecute	Prosecuted for Another Offense	Complaint Refused by D.A.	Prosecuted by Outside Department	Reprimanded and <u>Released</u>	Turned Over to Juvenile Authority	D.A. <u>Citation</u>	<u>Total</u>
Cleared										1
Count	0	43	Ō	2	0	11	1	46	1	104
% of row	0%	41.3%	0%	1.9%	0%	10.6%	1.0%	44.2%	1.0%	
% of column	Ō	100.0	0	100.0	0	100.0	100.0	100.0	100.0	
% of total	0	3.6	0	0.2	0	0.9	0,1	3.9	0.1	8.8%
Cleared-Other										
Count	0	0	34	Ō	4	0	0	0	0	38
% of row	0%	0%	89.5%	0%	10.5%	0%	0%	0%	0%	
% of column	0	0	91.9	0	100.0	0	0	0	0	
% of total	0	0	2.9	0	0.3	0	Q	Ò	Ó	3.2%
Uncleared										
Count	1042	0	3	0	Ó	Ō	0	0	. 0	1045
% of row	99.7%	0%	0.3%	<u>0</u> %	0%	0%	0%	0%	0%	
% of column	100.0	Ō	8,1	0	0	0	0	0	0	
% of total	87.8	<u> </u>	0.3	0	0	0	0	<u></u>	Ō	88.0%
Total Count	1042	43	37	2	4	11	1	46	. 1	1187
% of Cases	87.8%	3.6%	3.1%	0.2%	0.3%	0.9%	0.1%	3,9%	0.1%	100.0%

CHAPTER II. ROBBERY

As Table I-1 in Chapter I shows, robbery is the highest-volume crime committed against persons. For purposes of our analysis, we considered theft from person and purse snatch as forms of robbery, although they are classified as thefts by the FBI Uniform Crime Reports. We included these offenses because of the personal encounter that occurs between the victim and the offender, even though it is generally shorter than the encounter in an armed or strong-arm robbery.

After classifying robbery into the four subcategories, we then combined them into two groups:

	Group 1		Group 2							
			,							
•	Armed	•	Theft f	rom	person					
•	Strong-arm	•	Purse s	nato	:h					

Analysis of the frequencies of occurrence and levels of case clearances revealed that within each of the two groups the categories are similar. However, for cross-tabulation purposes, to determine the patterns of the incidents, we combined the two groups. When some cross tabulations produced results that appeared to be incompatible with the actual facts of how these subcategories of robberies were committed and cleared, we separated the cases and calculated their correlation coefficients by Groups 1 and 2 separately.



The analysis is presented in three sections:

- Selected cross tabulations on the entire robbery sample.
- Bivariate correlations and development of a decision rule using discriminant analysis, for the robbery cases in which there was CID involvement. (Our assumption was that CID involvement was likely to have occurred more than 8 hours after the report of the robbery.)
- Bivariate correlations and discussion regarding off-scene arrests made by patrol less than 8 hours after the report of the crime.

A. Cross Tabulations

In this section are selected tables of the sample cases processed, which illustrate the insight developed regarding the nature of robbery and the elements of information found to contribute to suspect ID and successful case closure. Further, the data reveal that patrol operational involvement in responding to robbery incidents has a profound impact of case closure at the patrol and investigation levels. Because showing hundreds of cross tabulations would not be productive, we illustrate only the most interesting analyses.

Analysis of Table II-1, Clearance by Primary Felony Offense, indicates that higher clearance rates were attained for the more serious categories of robbery--strong-arm and armed--than for theft from Person and purse snatch. This fact formed a major Part of our rationale for the grouping of robbery cases that has been shown. Table II-1 shows that approximately 15% of all robbery cases were cleared (cleared and cleared-other). Of the 818 cases of robbery, 5.7% were cleared strong arm robberies; 6.2% cleared armed robberies; 2.0% cleared thefts from person; and 1.3% cleared purse snatches. Of the 103 cases of the ft



Tabl. 11-1

ROBBERY: CLEARANCE BY PRIMARY FELONY OFFENSE

Status	Strong-arm Robbery			Purse Snatch	Total
Cleared	•			•	
Count	36	42	13	10	101
% of row	35.6%	41.6%	12.9%	9.9%	
% of column	13.1	12.7	11.8	9.7	· ·
% of total	4.4	5.1	1.6	1.2	12.3%
Cleared-Other				•	
Count	11	9	3	0	23
% of row	47.8%	39.1%	13.0%	0%	
% of column	4.0	2.7	2.7	0	
% of total	1.3	1.1	0.4	ο.	2.8%
Uncleared					
Count	228	279	94	93	694
% of row	32.9%	40.2%	13.5%	13.4%	
% of column	82.9	84.5	85.5	90.3	
% of total	27.9	34.1	11.5	11.4	84.8%
Total Count	275	330	110	103	8.8
% of Cases	33.6%	40.3%	13.4%	12.6%	100.0%

from person, 10 were cleared, and no cleared-other were shown. Because this category of street crime is truly a stranger-to-stranger crime, the chance of the victim knowing the offender and refusing to Press charges was nonexistent for the sample drawn.

Some weapon was used in 42.6% of the robbery cases. Table II-2 shows the weapons used. The two most common weapons were handguns, which were used in 18.7% of the cases, and knives, which were used in 7.7% of the cases.



Table II-2
ROBBERY: CLEARANCE BY WEAPON USED

Status	No <u>Weapon</u>	Håndgun	Rifle	Shotgun	Alleged <u>Gun</u>	<u>Knife</u>	Chemical	Blunt In- strument	Simulated Weapon	Other Weapon	Unknown Weapon	<u>Total</u>
Cleared		ŧ										
Count	51	12	Ō	3	3	10	0	ā				
% of row	50.5%	11.9%	0%	3.0%	3.0%	9.9%	0%	3 2 A#	8	11	0	101
% of column	10.7	7.8	0	60.0	21.8	15.8	0.6	3.0% 7.5	7.9%	10.9%	0%	
% of total	6.2	1.5	Ō	0.4	0.4	1.2	0	0.4	29.0	54.7	0	
Cleared-Other						- 1 =	٧	Ų. 4	1.0	1.3	0	12.3%
Count % of row	13	4	Ō	1	0	Ĵ	Q	1	Õ	1	0	23
% of row % of column	56.5%		4.3%	0%		0%	4.3%	9.70	4.3%	0%	<i>L</i>)	
% of total	2.7		0	20.0	0	4.7	0	2.5	0	5.0	0.0	
w at fâf e t	1.6	0.5	0	0.1	0	0.4	0	0.1	0	0.1	Õ	2.8%
Uncleared											2	£ (V /a
Count % of row % of column	414 59.6% 86.6	137 19.8% 89.6	9 1.3% 100.0	l 0.1% 20.0	11 1.6% 78.2	50 7.2% 79.5	4 0.6%	36 5.2%	20 2.8%	8 1.2%	4 0.6%	694
% of total	50,6	16.8	1.1	_0.l			100.0	90.0		40.3	100.0	
	· · · · · · · ·	-			1.3	<u>6.1</u> ,	0.5	4.4	2.4	<u>l.</u> 0	0.5	84.8%
Total Count	478	153	9	5	14	63	4	40	28	20	4	818
% of Cases	58.4%	18.7%	1.1%	0.6%	1.7%	7.7%	0.5%	4.9%	3.4%	2.5%	0.5%	100.0%

Table II-3 shows the importance of timely reporting of a robbery incident. In over 77% of the cases cleared by arrest, the report had been made within 2 hours of the crime's occurrence. The greatest percentage--65%--of the arrests occurred within the first hour.

It can readily be seen from Table II-4 that, of the 101 cases classified as cleared, 50% were cleared by arrest within the first 2 hours of the report of the incident. The overwhelming majority of these cases were cleared within 1 hour. The inference is that patrol is accounting for the largest percentage of robbery case clearance. Where case clearances are shown distributed over the indicated extended time intervals, e.g., longer than 8 hours of delay in arrest of a suspect, we assume that investigators were following up on leads provided by the initial reporting officers.

A further analysis of the time between report and arrest was undertaken for each of the four categories of robbery. (See Tables II-5 through -8).

In comparing the percentages of the cases cleared within 8 hours and of those whose clearance required more than 8 hours, it can be seen that, except for armed robbery, the greater percentages of the cases were those cleared within 8 hours. A summary comparison follows. [The parentheses (+) indicate the higher percentage and (-) the lower percentage.]

Strong-Arm Robbery	Theft Armed Robbery from Person Purse Snatch								
<8 hr 9.8% (+)	<8 hr 4.5% (-)	<8 hr 7.3% (+)	<8 hr 6.8% (+)						
>8 hr 4.0% (-)	>8 hr 6.0% (+)	>8 hr 3.6% (-)	>8 hr 3.0% (-)						

A probable deduction from the above tabulation is that CID investigators are more involved in clearing armed robbery cases than is patrol, but



Table II=3

ROBBERY: CLEARANCE BY TIME BETWEEN OCCURRENCE AND REPORT

	Status	Unknown Time	1 Hour	l to 2 Hours	2 to 4 Hours	4 to 8 Hours	8 to 12	12 to 24 Hours	l to 2 Days	2 to 4 <u>Daya</u>	4 to 7 Days	14 to 21 Days	30 to 45 <u>Days</u>	45 Plus _Days	Total
	Cleared Count % of row	1	66	12	5	0	2	Š	م	1	· Õ	1	1	i	
	% of column % of total	1.0% 3.2 0.1	65.3% 11.2 8.1	11.9% 14.0 1.5	5.0% 18.5 0.6	0% 0 0	2.0% 18.6 0.2	7.9% 25.7 1.0	3.0% 15.9 0.4	I.0% 13.3 0.1	0% 0	1.0%	1.0% 21.0	1 1.0% 100.0	101
N N	Cleared-Other Count	Ō	14	ĵ	2	Õ					Ō	Ō. <u>Ì</u>	0.1	0.1	12.3%
Ŋ	% of row % of column % of total	0% 0	60.9% 2.4	13.0% 3.5	8.7% 7.4	0% 0% 0	0 0% 0	2 8.7% 6.4	2 8.7% 10.6	0 0% 0	0 0% 0	0 07, 0	0 C% 0	0 0% 0	23
	Uncleared	0	1,7	0.4	0.2	0	• 0	0.2	0.2	0	0	0	0	0	2.8%
	Count % of row % of column % of total	31 4.47, 96.8 3.7	508 73.1% 86.4 62.0	71 10.2% 82.5 8.6	20 2.9% 74.1 	10 1.5% 100.0 _1.3	9 1.3% 81.4	21 3.0% 67.8	14 2.0% 73.4	7 0.9% 86.7	l 0.1% 100.0	0 0% 0	4 0.5% 79.0	0 0% 0	594
	Total Count	32	588	86	27	10	<u>l,l</u> ll	<u>2.6</u> 31	1.7	<u>0.8</u> 8	0.1	<u>)</u>	0.5	<u> 0 </u>	84.8%
f I	% of Cases	3.9%	71.8%	10.5%	3.3%	1.3%	1.3%	3.8%	2.3%	0.9%	0.1%	1 0.1%	5 0.6%	1 0.1%	818 100.0%

Table II-4

ROBBERY: CLEARANCE BY TIME BETWEEN REPORT AND ARREST

Status	No Arrest	Within <u>I Hour</u>	l to 2 Hours	2 to 4 Hours	8 to 12 Hours	12 to 24 <u>Hours</u>	l to 2 Days	2 to 4 Days	4 to 7 Days	7 to 10 <u>Days</u>	10 to 14 Daya	14 to 21 Days	21 to 30 Days	45 Plus Days	<u>Total</u>
Cleared															
Count	10	47	4	j	2	6	6	6	a, and	5	3	1	ļ	4	101
% of row	9.9%	46.5%	4.0%	3.0%	2.0%	5.9%	5.9%	5.9%	3.0%	5.0%	3.0%	1.0%	1.0%	4.07	
% of column	1.4	94.0	100.0	100.0	100.0	10 <u>0</u> .0	100.0	100.0	100.0	83.3	100.0	100.0	100.0	100.0	
% of total	1.2	5.7	0.5	0.4	0.2	0.7	0.7	0.7	0,4	0.6	0.4	0.1	0.1	0.5	12.3%
Cleared=Other															
Count	19	3	0	0	0	Ŋ	Ó	Ō	0	ļ	Ó	0	Ŏ	Ò	23
% of row	82.6%	13.0%	07.	07.	0%	0%	07,	0%	0%	4.3%	07.	07,	0%	0%	
% of column	2.5	6.0	Ō	Q	0	Ů	0	Ó	Ó	16.7	Õ	0	Ō	0	
% of total	2.3	0.4	Q	0	0	0	Ō	0	0	0.1	0	0	0	Ō	2.8%
Uncleared															
Count	694	0	0	Ó	Ō	Õ	Q	Ō	0	Ō	0	0	0	Ò	694
% of row	100.0%	0%	07,	07,	07,	Ò7 ,	0%	0%	0%	0%	0%	07.	0%	07.	
% of column	96.0	Õ	0	Õ	0	Ō	0	0	0	Ó	0	0	0	0	
% of total	84.8				<u>()</u>	<u> </u>	0	Ō		<u> </u>		0		0	84.87
Total Count	723	50	4	3	2	6	6	6	Ĵ	6	,	i	ļ	4	818
% of Cases	88.4%	6.1%	0.5%	0.4%	0.2%	0.7%	0.7%	0.7%	0.4%	0.7%	0.4%	0.1%	0.1%	0.5%	100.07.

Table II-5

ROBBERY, STRONG-ARM: CLEARANCE BY TIME BETWEEN REPORT AND ARREST

Status	No <u>Arrest</u>	Within 1 Hour	l to 2	2 to 4 Hours	8 to 12 Hours	12 to 24 Hours	l to 2 Days	2 to 4 Days	7 to 10 <u>Days</u>	14 to 21 Days	45 Plus Days	Total
Cleared												
Count	1	20	2	2	l	2	3	ļ	l	1	2	36
% of row	2.8%	55.6%	5,6%	5.6%	2.8%	5.6%	8.3%	2.8%	2.8%	2.8%	5.6%	
% of column	0.4	87.0	100.0	100.0	100.0	100.0	100.0	100.0	100.0	100.0	100.0	
% of total	0.4	7.3	0.7	0.7	0.4	0.7	1.1	0.4	0.4	0.4	0.7	13.1%
Cleared-Other												
Count	8	3	Õ	0	Ō	Ŏ	Ó	0	Ó	0	0	11
% of row	72.7%	27.3%	0%	0%	0%	0%	0%	0%	0%	Ō ⁵ / _e	0%	
% of column	3.4	13.0	Ō	0	Q	Ŏ	Ô	Ô	Ó	0	. 0	
% of total	2.9	1.1	Ō	Ō	0	0	0	0	0.	0	. 0	4.0%
Uncleared												
Count	228	Ō	0	0	0	0	0	0	0	Q	Q	228
% of row	100.0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	
% of column	96.2	Õ	Ó	0	0	0	. 0	Ō	0	Ō	Ō	
% of total	82.9	0	0	0	0	<u> </u>	<u> </u>	Ò			0_	82.9%
Total Count	237	23	2	2	i	2	Ĵ	Ţ	1	Ī	2	275
% of Cases	86.2%	8.4%	0.7%	0.7%	0.4%	0.7%	1.1%	0.4%	0.4%	0.4%	0.7%	100.0%

Table II-6

ROBBERY, ARMED: CLEARANCE BY TIME BETWEEN REPORT AND ARREST

	Status	No <u>Arrest</u>	Within 1 Hour	l to 2 <u>Hours</u>	2 to 4 Hours	8 to 12 Hours	12 to 24 Hours	l to 2 Days	2 to 4 Days	4 to 7 Days	7 to 10 <u>Days</u>	10 to 14 <u>Days</u>	21 to 30 <u>Days</u>	45 Plus <u>Days</u>	<u>Total</u>
	Cleared														
	Count	8	12	2	1	1	2	3	4	2	2	3	l	l	42
	% of row	19.0%	28.6%	4.8%	2:4%	2.4%	4.8%	7.1%	9.5%	4.8%	4.8%	7.1%	2.4%	2.4%	
	% of column	2.7	100.0	100.0	100.0	100.0	100.0	100.0	100.0	100.0	66.7	100.0	100.0	100.0	
	% of total	2.4	3.6	0.6	0.3	0.3	0.6	0.9	1.2	0.6	0.6	0.9	0.3	0.3	12.7%
	Cleared-Other							ı							
	Count	8	0	0	0	0	Ō	0	Û	Q	<u>İ</u>	0	Q	Ō	9
	% of row	88.9%	07,	0%	0%	0%	0%	0%	0%	0%	11.1%	0%	0%	0%	
	% of column	2.7	0	0	0	0	Ď	0	Ō	0	33.3	Q	Ō	Ō	
	% of total	2.4	0	Q	0	0	0	0	0	0	0.3	0	0	0	2.7%
	Uncleared														
•	Count	279	0	0	0	0	Ó	0	0	0	0	0	0	0	279
	% of row	100.0%	0%	٥%	0%	0%	0%	0%	0%	07.	0%	0%	0%	07,	
	% of column	94.6	0	٥	0	0	Ó	Ō	Ø	Ō	Q	Q	0	Õ	
	% of total	84.5	0	0	0	0				0	0	Ō	0	0	84.5%
	Total Count	295	12	2	1	1	2	3	4	2	3	3	1	1	330
	% of Cases	89.4%	3.6%	0.6%	0.3%	0.3%	0.6%	0.9%	1.2%	0.6%	0.9%	0.9%	0.3%	0.3%	100.0%

Table II-7

ROBBERY, THEFT FROM PERSON:
CLEARANCE BY TIME BETWEEN REPORT AND ARREST

Status	No Arrest	Within 1 Hou <u>r</u>	12 to 24 Hours	2 to 4 Days	7 to 10 Days	45 Plus Days	Total
Status	Allest	1 HOUL	Hours				
Cleared							
Count	1	8	1	1	1	1	13
% of row	7.7%	61.5%	7.7%	7.7%	7.7%	7.7%	
% of column	1.0	100.0	100.0	100.0	100.0	100.0	
% of total	0.9	7.3	0.9	0.9	0.9	0.9	11.8%
Cleared-Other							
Count	3	0	0	0	0	О	3
% of row	100.0%	0%	0%	0%	0%	0%	
% of column	3.1	0	0	0	0	0	
% of total	2.7	0	0	0	0	0	2.7%
Uncleared							
Count	94	0	0	0	О	0	94
% of row	100.0%	0%	0%	0%	0%	0%	
% of column	95.5	0	0	0	0	0	
% of total	85.5	0_	0	0	0	0	85.5%
Total Count	98	8	1.	1	1	1	110
% of Cases	89.1%	7.3%	0.9%	0.9%	0.9%	0.9%	100.0%

Table II-8

ROBBERY, PURSE SNATCH:

CLEARANCE BY TIME BETWEEN REPORT AND ARREST

Status	No <u>Arrest</u>	Within 1 Hour	12 to 24 Hours	4 to 7 Days	7 to 10 Days	Total
Cleared						
Count	0	7	1	1	1	10
% of row	0%	70.0%	10.0%	10.0%	10.0%	
% of column	0	100.0	100.0	100.0	100.0	
% of total	0	6.8	1.0	1.0	1.0	9.7%
Uncleared	•					
Count	93	0	0	0	0	93
% of row	100.0%	0%	0%	0%	0%	
% of column	100.0	0	0	0	0	
% of total	90.3	0	0_	0_	0	90.3%
Total Count	93	7	1	1	1	103
% of Cases	90.3%	6.8%	1.0%	1.0%	1.0%	100.0%



. . .

that patrol is more effective in the three other categories of robbery. Note that 29 clearances are shown for which no arrest was made. Usually this implies either that a warrant was issued or that the complainant refused to prosecute (Table II-4).

Tables II-9 through -11 show the races of the victims and offenders broken down into the cleared, cleared-other, and uncleared cases. These tables show that:

- Where white offenders committed robberies (all categories)
 against white victims, the overall clearance rate was 26%
 (11 clearances for a total of 43 cases).
- For black offenders with black victims, the clearance rate was 22% (45 clearances for a total of 203 cases).
- For black offenders with white victims, the clearance rate was only 11% (47 clearances for a total of 443 cases).

Table II-12 shows the races of the victims and offenders in the cleared and cleared-other cases where the suspect was known to the victim. The analysis was made to ascertain whether differences could be observed in the degree to which victims and suspects of different ethnic groupings were known to one another. The findings are that in the cleared cases:

- White offender/white victim--9% of the victims knew the offenders (1 cleared and known out of a total of 11 cleared cases).
- White offender/black victim--50% of the black victims knew the white offenders (1 cleared and known out of 2 cleared cases).



Table II-9

ROBBERY: VICTIM'S RACE BY OFFENDER'S RACE-CLEARED CASES

1		0 f	fender		
	Not				•
Victim	Known	White	Black	Mexican	Total
Not Known					
Count	0	0	2	0	2
% of row	0%	0%	100.0%	0%	
% of column	0	0	2.4	0	
% of total	0	0	2.0	0	2.0%
White					
Count	0	9	43	1	53
% of row	, 0%	17.0%	81.1%	1.9%	
% of column	0	75.0	50.6	33.3	
% of total	0	8.9	42.6 [.]	1.0	52.5%
Black					
Count	1	2	30	1	34
% of row	2.9%	5.9%	88.2%	2.9%	
% of column	100.0	16.7	35.3	33.3	
% of total	1.0	2.0	29.7	1.0	33.7%
Mexican					
Count	0	0	4	1	5
% of row	0%	0%	80.0%	20.0%	
% of column	0	0	4.7	33.3	
% of total	0	0	4.0	1.0	5.0%
Japanese					
Count	0	0	2	0	2
% of row	0%	0%	100.0%	0%	
% of column	0	0	2.4	0	
% of total	0	0	2.0	0	2.0%
Other					
Count	0	1	4	0	5
% of row	0%	20.0%	80.0%	0%	
% of column	0	8.3	4.7	0	
% of total	0	1.0	4.0	0	5.0%
Total Count	1	12	85	3	101
% of Cases	1.0%	11.9%	84.2%	3.0%	100.0%



Table II-10

ROBBERY: VICTIM'S RACE BY OFFENDER'S RACE-CLEARED-OTHER CASES

		Offender	<u>r </u>	
Victim	White -	Black	<u>Mexican</u>	Total
White	•			
Count	2	4	0	6
% of row	33.3%	66.7%	0%	
% of column	66.7	21.1	0	
% of total	8.7	17.4	0	26.1%
Black	/			•
Count	0	15	0	·15
% of row	0%	100.0%	0%	
% of column	0	78.9	0	
% of total	0	65.2	0	65.2%
Mexican			r	
Count	0	. 0	1	1
% of row	0%	0%	100.0%	
% of column	0	0	100.0	
% of total	0	0	4.3	4.3%
Japanese				
Count	1	0	0	1
% of row	100.0%	0%	0%	
% of column	33.3	0	0	
% of total	4.3	0	0	4.3%
Total Count	3	19	1	23
% of Cases	13.0%	82.6%	4.3%	100.0%



Table II-11

ROBBERY: VICTIM'S RACE BY OFFENDER'S RACE-UNCLEARED CASES

		Of	fender	···	
Victim	Not <u>Known</u>	.White	Black	Mexican	Total
Not Known					
Count ·	1	4	3	0	8
% of row	11.9%	52.3%	35.8%	0%	Ů
% of column	3.8	11.3	0.5	0	
% of total	0.1	0.6	0.4	Ö	1.2%
White					
Count	9	32	396	13	449
% of row	1.9%	7.1%	88.2%	2.9%	
% of column	32.2	81.7	64.9	66.2	
% total	1.2	4.6	57.0	1.9	64.7%
Black				•	
Count	11	3	158	0	172
% of row	6.6%	1.6%	91.8%	0%	
% of column	43.0	7.0	25.9	0	
% of total	1.6	0.4	22.7	0	24.8%
Mexican					
Count	6	0	20	7	32
% of row	17.2%	0%	62.3%	20.5%	
% of column	21.0	0	3.3	33.8	
% of total	0.8	0.	2.9	1.0	4.6%
American Indian				ý.	
Count	0	0	4	0	4
% of row	0%	0%	100.0%	0%	
% of column	0	. 0	0.7	0	
% of total	0	. 0	0.6	0	0.6%
Chinese	·		an are	W. L. L. 425 - 444 - 1	11 - 200
Count	0	0	10	0	10.
% of row	0%	0%	100.0%	0%	
% of column	0	. 0	1.7	ο,	
% of total	. 0	0	1.5	0	1.5%
Other					
Count	0	o _,	18	0	18
% of row	0%	0%	100.0%	0%	
% of column	0	0	2.9	0	
% of total		0	2.6	0	2.6%
Total Count	26	39	609	20	694
% of Cases	3.8%	5.6%	87.8%	2.8%	100.0%



Table II-12

ROBBERY: VICTIM'S RACE BY OFFENDER'S RACE-OFFENDER KNOWN TO VICTIM
(Cleared and Cleared-Other Cases)

Constitution

		Offende	r		
Victim	White	Black	Mexican	Total_	
White				•	
Count	1	4	. 0	5	
% of row	20.0%	80.0%	0%		
% of column	50.0	14.3	0		
% of total	3.1	12.5	0	15.6%	
Black					
Count	1	24	1	26	
% of row	3.8%	92.3%	3.8%		
% of column	50.0	85.7	50.0	•	
% o f total	3.1	75.0	3.1	81.3%	
Mexican					
Count	0	0	1	1	
% of row	. 0%	0%	100.0%		
% of column	0	0	50.0		
% of total	0	0	<u>3.1</u>	3.1%	
Total Count	2	28	2	32	
% of Cases	6.3%	87.5%	6.3%	100.0%	

- Black offender/white victim--9% of the white victims knew the black offenders (4 cleared and known out of 47 cleared cases).
- Black offender/black victim--53% of the black victims knew the black offenders (24 cleared and known out of 45 cleared cases).

The deduction from the above is that among all the robbery categories the probability that the victim can name the perpetrator is highest when both the victim and the offender are black.

B. Bivariate Correlations and Decision Models

The next step in our analysis of the robbery data was to analyze the cleared cases where there had been CID input. As stated earlier, we put into this group the cases classified as cleared or cleared-other where an arrest had not been made within 8 hours from the time of report of the crime. The cleared and cleared-other cases were considered together, because bivariate correlations run separately with the two groups indicated that their correlations with the variables under consideration differed only slightly. In addition, the larger sample size increased the statistical significance of the analysis.

From our examination of the cross tabulations and subjective interpretation of the data, we chose 108 variables for further analysis and potential inclusion in the decision model. (These are listed in Appendix D, Table D-1.) Bivariate correlations were run with these variables for both the strong-arm/armed robbery and theft from person/purse snatch groups. The variables that showed at least 0.1 correlation with clearance are listed in Tables II-13 and -14. These tables show extensive overlap between the two categories of robbery, especially among the



STRONG-ARM/ARMED ROBBERY VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS OF CLEARANCES REQUIRING MORE THAN 8 HOURS

	Correlation
Variable	Coefficient
Suspect named (TP/PS)*	0.4621
Suspect known (TP/PS)	0.4365
Suspect previously seen (TP/PS)	0.4066
License number of vehicle given	0.2889
Vehicle registration checkuseful lead	
(TP/PS)	0.2848
Field Contact report	0.2570
Places suspect frequented named (TP/PS)	0.2480
Other physical evidence match	0.2202
Offender and victim same race (TP/PS)	0.2106
Evidence technician at crime scene	0.2072
Three or more reporting individuals	0.2047
White offender	0.2036
Suspect's associates named/indicated (TP/PS)	0.2014
Greater than 30 minutes contact between	
victim and offender (TP/PS)	0.1959
White offender and black victim (TP/PS)	0.1797
Crime File run-personuseful lead (TP/PS)	0.1797
Clothing match	0.1797
Weapons match	0.1797
Black victim (TP/PS)	0.1783
Description of vehicle given	0.1701
Vehicle registration check made (TP/PS)	0.1639
Crime occurrence between 0801 and 1200 hours	0.1630
Fingerprints taken	0.1554
Vehicle registered to suspect (TP/PS)	0.1551
Vehicle used	0.1500
Color of vehicle given	0.1484
Other weapon used (TP/PS)	0.1483
Black offender and black victim	0.1471
Victim invited offender in (TP/PS)	0.1460
Sexual aberrations indicated	0.1389
White offender and white victim	0.1382
Offender movement by automobile	0.1314
Fingerprints match	0.1269
Offender described as wearing glasses 🖑	0.1112
One offender	0.1017
•	

^{*}TP/PS--Also significant for theft from person/purse snatch.





Table II-14

THEFT-FROM-PERSON/PURSE-SNATCH ROBBERY VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS OF CLEARANCES REQUIRING MORE THAN 8 HOURS

Variable	Correlation Coefficient
Vehicle registration checkuseful lead (S/A)*	0.4165
Offender silentnote passed	0.4165
Suspect named (S/A)	0.4046
Suspect known (S/A)	0.3656
Suspect previously seen (S/A)	0.3469
Words spoken by offender	0.3235
Crime File run-personuseful lead (S/A)	0.2938
Offender violent	0.2938
Greater than 30 minutes contact between	
victim and offender (S/A)	0.2755
Offender pretended	0.2242
Black offender and black victim (S/A)	0.2186
Crime occurred between 0401 and 0800 hours	0.2148
Places suspect frequented named (S/A)	0.2105
Vehicle registered to suspect (S/A)	0.1797
Three clothing descriptors given	0. 1633
Black victim (S/A)	0.1557
Victim invited offender in (S/A)	0.1461
Two reporting individuals	0.1345
Eyes of offender described	0.1295
Other weapon used (S/A)	0.1292
Crime file run-vehicle	0.1292
Offender and victim same race (S/A)	0.1283
Suspect's associates named/indicated (S/A)	0.1258
Female offender	0.1217
Mexican-American offender	0.1213
Vehicle registration check made (S/A)	0.1118

 $^{^{*}}$ S/A--Also significant for strong-arm and armed robbery.



variables with the highest correlations with clearance. We therefore decided to construct one decision rule for all categories of robbery.

We selected the data elements for further screening on the basis of their correlation coefficients and our subjective judgment of the usefulness of certain data elements for police investigative purposes. For example, we initially assumed that variables concerned with weather and illumination (rain, fog, clear, daylight, dawn, dusk, dark, artificial light) would be reported by the beat officer when interrogating a victim of a street crime, to determine factors affecting the victim's ability to describe an offender. In the 818 cases sampled, one of the variables—dark—was noted only seven times. The others were noted three times or usually not mentioned. The SRI data coders could have made assumptions about the state of darkness by noting the time of a crime, but the reports usually did not mention street illumination. Consequently, we eliminated these variables from further consideration.

Not listed in Table II-13 are some variables that we fully expected to have some statistical significance. Elements of information on suspect physical descriptions, such as height, weight, eyes, hair, glasses, and teeth, all exhibited negative correlation coefficients for case clearances by arrest or were below the 0.1000 threshold level established. Use of handguns also showed negative correlation with clearance. There are logical explanations for the behavior of these variables. Practically every report of an incident contains some of these descriptions. But the fact that most cases are uncleared, even though some of these decriptors appeared in both cleared and uncleared cases, indicates that the physical descriptor elements are not prime suspect identifiers. The negative correlation reveals that more uncleared cases contained this variable, e.g., handgun, than did cleared cases.



A similar explanation holds for wearing apparel descriptors having little prime impact on suspect ID. However, note that the variable clothing match in Table II-13 shows a contribution to case clearance. Again there is a logical explanation. Suppose that a suspect was apprehended on the basis of some other information lead in the set shown in Table II-13. If a victim or witness had described the offender's apparel, the description had been recorded in the report, and a suspect had been apprehended wearing the garments recorded, these procedures would have supported a positive ID. This variable then adds weight to suspect ID and case clearance by arrest.

Before a discriminant analysis could be successfully undertaken, however, it was necessary to restructure many of the variables to ensure that they were independent from one another (as discussed in Appendix D). Variables that were restructured included: vehicle, vehicle registration check, and physical evidence variables.

Many iterative discriminant analyses were run using various combinations of variables with varying methods of restructuring. This multistaged procedure was necessary to establish the set best able to discriminate between the cleared and uncleared cases, and to predict with a high degree of accuracy the group (cleared or uncleared) to which a particular case belonged. The analysis below describes some of the major decisions we made leading to the development of the decision model.

An important (perhaps obvious) decision was to exclude cases that had been solved on the basis of the suspect being named or known. The police are the first to point out that, if the suspect is identified in the crime report, the case is essentially solved. The statistics supported this observation. When these two variables are included in discriminant analysis, their presence is so dominating that the other variables seem worthless.

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A second reason for excluding cases where the suspect had been named or known is that we were trying to gain insight into what other investigative leads are important in case clearance. A case where the suspect is known requires little investigation except to develop the case for prosecution. A bivariate correlation analysis on the four robbery subgroups, of cases that required more than 8 hours between report and clearance, with the offender neither named nor known, produced the list of correlated variables shown in Table II-15.

Table II-15

VARIABLES FOR CASES WITH CLEARANCES
REQUIRING MORE THAN 8 HOURS
AND OFFENDER UNNAMED AND UNKNOWN
USED IN THE DISCRIMINANT ANALYSIS

	Correlation
Variable	Coefficient
Suspect previously seen	0.3410
Total physical evidence matched	0.3243
Evidence technician	0.2979
Places suspect frequented named	0.2858
Vehicle registration check	0.2398
Sexual aberrations	0.2138
Offender movement description	0.1822
Duration of contactvictim/offender	0.1607
Weapon used	0.1276
Offender/victim race	0.1193
Number of reporting individuals	0.1084
Total number of physical descriptors	0.1052
Total cash value of property taken	0.1030

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Table II-16 shows the results of the discriminant analysis performed on the selected set of 13 variables. The standardized discriminant function coefficients on the right of the table provide the ranking of importance to case clearance.

Table II-16
ROBBERY DISCRIMINANT FUNCTION COEFFICIENTS

Variable	Discriminant Function Coefficients
Suspect previously seen	0.65
Evidence technician	0.61
Places suspect frequented named	0.42
Total physical evidence matched	0.37
Vehicle registration information	0.22
Offender movement description	0.21
Weapon used	0.16
Offender/victim race	0.12
Total cash value of property	0.11
Total physical descriptors	0.08
Number of reporting individuals	0.03
Duration of contact	0.04
Sexual aberration	-0.08

On the basis of the discriminant function coefficients calculated, we selected seven variables to be used in the decision model. In addition, the variables suspect named and suspect known are included, but set apart from the other categories. Table II-17 displays the robbery investigation decision model constructed on a relative scale of 10. The weighted variables in the model reflect the contribution of the element of information to prediction of case clearance. The importance of each item is relative to that of all the other elements.



Table II-17 ROBBERY INVESTIGATION DECISION MODEL

Information Element	Weighting Factor
Suspect named	10*
Suspect known	10*
Suspect previously seen	10*
Evidence technician used	10
Places suspect frequented named	10*
Physical evidenceeach item matched	6.1
Vehicle registration	
Query information available	1.5
Vehicle stolen	3.0
Useful information returned	4.5
Vehicle registered to suspect	6.0
Offender movement description	
On foot	0
Vehicle (not car)	0.6
Car	1.2
Car color given	1.8
Car description given	2.4
Car license given	3.0
Weapon used	1.6

INSTRUCTIONS

- (1) Circle the weighting factor for each information element that is present in the incident report.
- (2) Total the circled factors.
- (3) If the sum is less than 10, suspend the case; otherwise, follow up the case.
- (4) Weighting factors do not accumulate; i.e., if both the auto license and color are given, the total is 3.0 not 4.8.



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These values as calculated actually exceed the threshold of 10. The values provided here are conceptually simpler and make no difference in the classification of groups.

The reader must always keep in mind that the decision model shown in Table II-17 was based on the OPD's operational practices that affect case handling, and consequently the manner by which cases are cleared. Other departments may have different policies, procedures, and capabilities. Consequently, the decision model usage must be carefully considered in light of each agency's operational procedures. It can be seen that our decision model contains a number of items of information resulting from preliminary enrichment procedures routinely performed by OPD personnel. This implies that the screening process should take place after certain basic investigative procedures, e.g., license number checks, have been made.

A further important consideration is that this model should be considered a dynamic model. The individual weighted categories of investigative information should be checked throughout the investigatory phase of the case. If a suspect is then not identified, the case can realistically be set aside as unsolvable.

A description of the variables in the decision model follows:

- Named and known. If the suspect is either named at the time
 of report or known to either the victim or a witness, the
 case is to be assigned a weight of 10 and therefore should
 be investigated.
- Suspect previously seen. If either the victim or a witness has previously seen the suspect, although this person is unable to name the suspect, the case should be pursued.
- Evidence technician. We rarely found that physical evidence had led to the initial identification of a suspect, although it did contribute to the strengthening of a case. However, the presence of an evidence technician at the crime scene



indicated an a priori judgment on the part of a patrol officer that physical evidence was present. Evidence technicians are a limited resource and are generally called to the scene only when the likelihood of clearance appears to be good.

- Places suspect frequented named. Presence of this liable indicated that either a witness or the victim, although unable to name the suspect, was able to provide information regarding where the suspect lived or worked, or places he or she frequented, e.g., bars.
- Offender movement description. A case was found to be more likely to be solved when an auto was involved, particularly when the license number was given.
- Total physical evidenced matched. We found no particular piece of physical evidence heavily contributory to clearance. Nor did we find that the mere presence of elements of unmatched physical evidence was associated with case clearance. The match variable implies that when a suspect has been apprehended, any physical evidence found on his person matching a description obtained from a crime scene, corroborates his identity as the offender.
- Weapon used. The use of a weapon in a robbery was found to contribute slightly to clearance of the case.
- Vehicle registration. If a license number is provided, a vehicle registration check is run, and the registered owner is identified. If the vehicle has been reported stolen, this is of some investigative value. However, if the vehicle is registered to the suspect, this is of greater investigative importance.



Ninety percent of the cases in our sample were correctly grouped as cleared or uncleared by the classification function derived from the discriminant analysis and reflected by the relative scaling in the decision model. The 10% error can be explained as follows: 8 cases that were eventually cleared were classified as uncleared, and 14 cases that remained uncleared were placed into the cleared category. The 8 cleared cases categorized as uncleared might initially seem to be a cause for concern. However, we pursued a further analysis to determine how these cases were eventually solved and whether an initial screening would have resulted in their not being cleared. Of the 8 cases, 6 were solved without investigation on the part of OPD detectives:

- Two were bank robberies investigated by the FBI (all bank robberies are referred to the FBI regardless of the information available).
- In 3 cases the offender was later linked to a robbery case when he was found in possession of the property reported as stolen.
- In one case, the suspect turned himself in.

In 2 cases, investigation by OPD detectives did lead to the eventual identification of a suspect, despite the fact that little information was available initially.

Table II-18 illustrates the case disposition decision rule developed for burglary follow-up screening in a prior SRI research project. The methodology for the development of the robbery model evolved from the earlier reported research. Attention is drawn to the similarity of the variables in the two rules and their relative weights in contributing to case clearances. In both models the victim of the crime, or a witness viewing it, provides the most useful information leading to case clearance.



Table II-18
BURGLARY CASE DISPOSITION DECISION RULE

Information Element	Weighting Factor
Estimated range of time of occurrence	
Less than 1 hour	5
1 to 12 hours	1
12 to 24 hours	0.3
More than 24 hours	0
Witness's report of offense	7
On-view report of offense	1
Usable fingerprints	7
Suspect information developed	
description or name	9
Vehicle description	0.1
Other	<u>o</u>
Total score	

INSTRUCTIONS

- (1) Circle the weighting factor for each information element that is present in the incident report.
- (2) Total the circled factors.
- (3) If the sum is less than or equal to 10, suspend the case; otherwise, follow up the case.

From B. Greenberg et al., "Enhancement of the Investigative Function," Vols. I and IV, Stanford Research Institute, Menlo Park, California (1972-1973).



But in the robbery cases, apart from the naming of the suspect, which also is a dominant element in burglary case clearance, vehicle information is the next-most-important information element leading to suspect ID.

The burglary model was validated by drawing additional case samples from selected participating agencies. The decision rule was used as a guide to select which cases would be cleared if followed up. We were somewhat dismayed to discover that a range of predictive accuracy in case selection varied from a high of 90% to a low of 67%. The explanation for the wide variation is simply that the agencies involved had inconsistent policies governing the criteria by which a burglary case is cleared. The highest accuracy was associated with an agency whose case clearance policies were extremely consistent with the evidence leading to suspect ID, arrest, and prosecution for the offense. The other agencies had less stringent policies.

The burglary model was independently evaluated by a study team in the OPD. The group drew a random sample of approximately 300 burglary cases, which were screened by a combination of personnel: analysts, clerks, and a police intern. The cases screened by use of the model numerical weighting scale were compared to cases actually selected by trained investigators for case follow-up or suspension.

The results of this comparison showed that the case scaling checklist methodology provided a more accurate besis on which to predict subsequent clearance. For example, in one experiment using an analyst and a clerical assistant, the checklist consistently predicted 71% of all clearances and 92% of arrests classified as cleared by arrest and prosecution. When clearances and investigations were compared, using the checklist and investigator for one mode of comparison, and the investigator only for a second mode, a clearance-to-investigation (CI) ratio of about 76% resulted in cases chosen for follow-up by both the



checklist and the investigators. In contrast, the CI ratio dropped to approximately 55% for the cases selected only by the investigators.

In the OPD experiment, both the checklist and the investigator modes of case selection produced small Type I and Type II errors, i.e., missing a case that was subsequently cleared or selecting a case or follow-up that was obviously cold. However, a number of cases that the checklist selected for follow-up were cases that the analyst thought should have been followed up but were not. In summary, the OPD experiment indicated that, for agencies having a large volume of burglary reports to handle (Oakland reported 14,000 for 1971), an appreciable amount of skilled investigator time could be spared by having a semiskilled clerk prescreen the burglary reports, using the case selection checklist methodology.*

It appears desirable that a similar series of validation tests should be conducted with the robbery decision rule. It should be noted that the predictive accuracy obtained by the OPD study was achieved on the basis of the burglary decision rule, which did not reflect the OPD case clearance policies. The surprisingly high accuracy obtained may be attributable to the OPD's policy of stringent case clearance criteria.

C. Elements of Information Associated with Patrol Case Clearances

Patrol is evidently accounting for the largest percentage of case clearance for robbery taken as a whole. Of the cleared cases, 57% were cleared in less than 8 hours, with 47% cleared within 1 hour. Our assumption is that after 8 hours there is some CID involvement. Beyond 8 hours, 41% of the cleared cases were cleared; we have no indication of the time involved for 2% of the cases (accounted for by cases in which

^{*}Greenberg, et al., op. cit., Vol. IV, pp. 10 through 15, and Appendix B.

warrants had been issued, and others where no arrests were made, but cases were cleared-other). (See Table II-4.) Although there would be little value to developing a decision rule for patrol in responding to a robbery incident, it is useful to examine, by the methodology described above, the categories of information contributing to the success of patrol in apprehending suspects.

Table II-19 lists, in order of decreasing correlation, the significant variables derived from the bivariate correlation analysis for variables associated with strong-arm/armed robbery cases where an off-scene arrest was made less than 8 hours after the crime had been reported.

Table II-19

STRONG-ARM/ARMED ROBBERY VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS OF OFF-SCENE ARRESTS OCCURRING IN LESS THAN 8 HOURS

	Correlation
Variable	<u>Coefficient</u>
Total number of physical evidence matches (TP/PS)*	0.5498 0.3541
Evidence collected and matched (TP/PS) Duration of contactvictim/offender (TP/PS)	0.2809
Total amount of physical evidence (TP/PS)	0.2214
Field contact reportuseful	0.2184
Offender movement	0.1922
Suspect previously seen	0.1902
Victim cooperative	0.1300
Number of reporting individuals (TP/PS)	0.1299
Suspect known	0.1180
Suspect named	0.1042
Occurrence between 0801 and 1200 hours (TP/PS)	0.1031

^{* (}TP/PS)--Also significant for theft from person/purse snatch.



An interesting observation about the variables listed in Table II-19 (showing significance in patrol clearance) is that those concerned with evidence and with matches of evidence appear to be the most important contributing factors to suspect ID. Clothing, race, and physical descriptors appear to contribute minimally to suspect ID in a statistical sense. We can conjecture as to the factors accounting for nonappearance of these elements in our data. The officer making the arrest may have had information on the offender's description, but neglected to record it on the report. We know from practical experience that clothing descriptors have a certain time usefulness in searching for a fleeing offender. But the fact that 508 robbery reports (out of 818 cases sampled) were reported within 1 hour of occurrence and were not cleared shows that whatever descriptions were provided did not contribute heavily to the overall case clearances.

The fact that the victim provided an indication of who the offender was or could name him appears to have been significant contributory factors to case clearance. In some areas, when the victim described certain pieces of evidence and these matched those in possession of the offender, a positive ID was made. This latter fact is borne out by the high correlation coefficients associated with physical evidence. Since we collected data from the official reports of incidents, our statistical results naturally reflect the information that the reporting and investigating officers recorded.

We also looked at case clearance factors associated with theft from person/purse snatch case for off-scene arrests occurring less than 8 hours after report. Table II-20 lists, in order of decreasing correlation, the variables derived from the bivariate correlation analysis.



Table II-20

THEFT-FROM-PERSON/PURSE-SNATCH ROBBERY VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS OF OFF-SCENE ARRESTS OCCURRING IN LESS THAN 8 HOURS

Variable	Correlation Coefficient
Total amount of physical evidence (S/A)*	0.6178
Total number of physical evidence matches (S/A)	0.6178
Evidence collected and matched (S/A)	0.6178
Offender silent	0.5018
Offender violent	0.5018
Words spoken	0.2442
Direction of flight provided	0.2101
Offender pretended to be	0.1764
Duration of contactvictim/offender (S/A)	0.1724
Occurrence between 0801 and 1200 hours (S/A)	0.1715
Places suspect frequented named	0.1637
Number of reporting individuals (S/A)	0.1543
Person attacked	0.1285
Height of offender given	0.1103
Black offender/white victim	0.1043
Victim cooperative	0.1022
Time between occurrence and report	0.1002

^{* (}S/A)--Also significant for strong-arm/armed robbery.

Table I-4 indicates that 13 cases of theft from person were cleared and 3 were cleared-other. There were 10 cleared purse snatch cases. Of the 13 total clearances for theft from person, 8 were cleared by arrest within 1 hour. Although the clearance percentages are high for the less-than-8-hour category--62% for theft from person and 70% for purse snatch--the overall clearance rate for these two categories is low: 14.5% and 9.7%, respectively. If we use Table II-3 as an indicator of the overall robbery clearance rate as a function of time between



occurrence and report, we conclude that nearly 72% of all robberies are reported within 1 hour, but only 13.6% are cleared. We can only conjecture from these data whether the clearance rate is coupled to the rapidity of patrol response or whether the quality of data greatly affects the successful apprehension of the suspect. Unfortunately, we could not capture a potentially useful piece of information: the time of arrival of a police officer at the scene. The incident report does not show this time factor. The complaint-dispatch card shows the time the complaint was received at the OPD Communications desk. But a major effort would be entailed to link this information to patrol response and time of arrival on scene.

In summary, the best is erence we can draw from the less-than-8-hour clearances is that Tables II-19 and -20 show that clearances are based on the victim's providing some indication of knowing the offender and then the patrol officer quickly responding to pick him up (within 1 hour) and finding him in possession of some form of identifiable physical evidence.



CHAPTER III. ASSAULT WITH A DEADLY WEAPON

Assault with a deadly weapon (ADW) differs from the other felonies analyzed in that most of the cases were cleared. As shown in Table I-7, 49.9% of the cases were classified cleared, and 35.6% were classified cleared-other, for a total clearance rate of 85.5%.

ADW was selected for analysis for two principal reasons:

- ADW is part of the larger category of felony assault, which
 is one of the Part I crimes versus persons. We decided to
 concentrate our efforts on ADW because the ADW felonies are
 by far the most numerous within the category of felony
 assault.*
- We were interested in investigating the criminal histories of persons suspected of committing ADWs to ascertain any previous involvement in other criminal activity.

In this section we first present a number of interesting cross tabulations prepared from our data and then discuss the investigative inferences regarding case clearance that can be drawn from the data.

The predominant characteristic of the ADW cases was that in 280 (69%) of the total of 413 cases the victims knew the offenders (Table III-1).

Table III-2 shows the races of the offenders and victims in the cases where they were known to each other. Persons of the same race were involved in 239 (85%) of the 280 cases; in 222 of these cases (79% of the



^{*}Felony assault also includes "assault with intent to murder," "shooting at dwelling," and "child or wife beating." In California, the penal code for ADW is P.C. ^45.

Table III-1
ADW: CLEARANCE BY SUSPECT KNOWN

Status	No t Known	Known to	Known to Witness	Known to	Known to Other Person	<u>Total</u>
Cleared						
Count	54	140	9	1	1	206
% of row	26.1%	68.2%	4.3%	0.7%	0.7%	
% of column	45.1	50.1	81.7	100.0	100.0	
% of total	13.0	34.0	2.2	0.4	0.4	49.9%
Cleared-Other						
Count	14	133	0	0	0	147
% of row	9.8%	90.2%	0%	0%	0%	±-T/
% of column	12.1	47.4	0	0	0	
% of total	3.5	32.1	0	0	0	35.6%
Uncleared				e		
Count	51	7	. 2	0	0	60
% of row	85.0%	11.7%	3.3%	0%	0%	3.2
% of column	42.8	2.5	18.3	0	0	
% of total	12.3	1.7	0.5	0	0	14.5%
Total Count	119	280	11.	1	1	413
% of Cases	28.8%	67.8%	2.6%	0.4%	0.4%	100.0%

280 cases) the victims and offenders were black. Table III-3 shows the races of the offenders and victims in the 133 cases where they were not known to each other. Persons of the same race were involved in 65 (49%) of the 133 cases. Thus, when a person was assaulted by a person of the same race, they were more likely to be known to each other than when the victim and offender were of different races.

Maria

Table III-4 shows the time of occurrence of the ADWs. As might be expected, they were concentrated in the late night and early morning hours, with 26.3% occurring between 8 p.m. and midnight, and another 21.2% between midnight and 4 a.m.



Table III-2

ADW: OFFENDER'S RACE BY VICTIM'S RACE-OFFENDER KNOWN TO VICTIM

			7	/ictim				
	Not				American			
Offender	Known	White	Black	Mexican	Indian	Japanese	Other	Total
			,					
Not known							_	
Count	0	0	2	0	0	О	0	2
% of row	0%	0%	100.0%	0%	0%	0%	0%	
% of column	0	0	1.1	0	0	0	0	s a#
% of total	0	0	0.9	0	0	0	0	0.9%
White								
Count	0	7	6	1	0	0	1	15
% of row	0%	46.5%	40.1%	6.7%	0%	0%	6.7%	
% of column	0	23.3	2.6	13.5	0	U	28.8	
% of total	0	2.5	2.1	0.4	0	0	0.4	5.3%
	-							
Black		1.0	222	1	0	0	1	245
Count	3	18	222		0%	0%	0.4%	
% of row	1.2%	7.4%	90.3%		0/3	0,	28.8	
% of column	66.7	61.6	96.3	19.9	0	o o	0.4	87.6%
% of total	1.1	6.5	79.2	0.5	Ü	Ü	. .	9,
Mexican							•	o
Count	0	3	0	5	0	0	0	8
% of row	0%	37.8%	0%		0%	0%	0%	
% of column	0	10.1	0	66.5	0	0	0	o 09/
% of total	0	1.1	0	1.8	0	0	0	2.8%
American Indian		,						
Count	1	1	0	0	3	0	0	6
% of row	23.1%	23.1%	0%	. 0%	53.9%	0%	0%	
% of column	33.3	5.0	0	0	100.0	0	0	
% of total	0.5	0.5	0	0	1.2	0	0	2.3%
Japanese								
Count	0	0	0	0	0	1	0	1
% of row	0%	0%	0%	ر ۱ 0%	0%	100.0%	0%	
% of column	0	0	0	0	0	100.0	0	
% of total	0	0	0	0	0	0.5	0	0.5%
0.1								
Other	0	0	0	0	O	0	. 1	1
Count	0%	0%	0%	_	0%	0%	100.0%	
% of row % of column	_0	0	0	0	0	0	42.5	
% of total	0	0	0		0	0	0.5	0.5%
				- 	3	1	3	280
Total Count	4	30	230					
% of Cases	1.6%	10.6%	82.2%	2.6%	1.2%	0.5%	1.2%	100.0%



ADW: OFFENDER'S RACE BY VICTIM'S RACE-OFFENDER NOT KNOWN TO VICTIM

Victim								
	Not				American			
Offender	Known	White	Black	Mexican	Indian	Total		
Not known								
Count	. 0	4	5	1	0	10		
% of row	0%	40.0%	50.0%	10.0%	0%	10		
% of column	0	7.7	8.5	9.6	0			
% of total	0	3.0	3.8	0.8	0	7.5%		
White								
Count	3	13	1	1	1	20		
% of row	14.9%	65.1%	7.5%	7.5%	5.1%			
% of column	27.3	24.9	2.5	14.2	100.0			
% of total	2.2	9.7	1.1	1.1	0.8	14.9%		
Black								
Count	6	32	50	6	0	95		
% of row	6.8%	34.1%	52.8%	6.3%	0%			
% of column	59.1	62.6	84.8	57.1	0			
% of total	4.8	24.3	37.6	4.5	0	71.3%		
Mexican								
Count	0	0	1	2	· 0	3		
% of row	0%	0%	42.5%	57.5%	0%			
% of column	0	0	2.5	19.2	0			
% of total	0	0	1.1	1.5	0	2.6%		
American Indian								
Count	1	1	0	0	0	3		
% of row	50.0%	50.0%	0%	0%	0%			
% of column	13.6	2.9	0	0	0			
% of total	1.1	1.1	0	0	0	2.2%		
Other								
Count	0	1	1	0	0	2		
% of row	0%	50.0%	50.0%	0%	0%			
% of column	0	1.9	1.7	0	0			
% of total	0	0.8	0.8	0	0	1.5%		
Total Count	11	52	59	10	1	133		
% of Cases	8.1%	38.9%	44.4%	7.8%	0.8%	100.0%		



Table III-4

ADW: CLEARANCE BY TIME OF OCCURRENCE

	Status	Unknown Time	0001-0400 Hours	0401-0800 Hours	0801-1200 Hours	1201-1600 Hours	1601-2000 Hours	2001-2400 Hours	<u>Total</u>
	Cleared								
	Count	0	34	7	19	47	42	57	206
	% of row	0%	16.3%	3.6%	9.1%	22.8%	20.4%	27.8%	200
	% of column	0	38.3	30.7	51.3	62.1	53.1	52.6	
	% of total	0	8.1	1.8	4.5	11.4	10.2	13.9	49.9%
	Cleared-Other								
Ųī Ųī	Count	0	35	14	15	23	25	35	147
Ŧ'	% of row	0%	23.8%	9.4%	10.1%	15.4%	17.1%	24.1%	±4 <i>1</i>
	% of column	0	40.0	57.0	40.5	30.0	31.8	32.6	
	% of total	0	8.5	3.3	3.6	5.5	6.1	8.6	35.6%
	Uncleared								
	Count	1	19	3	3	6	12	16	60
	% of row	1.7%	31.7%	5.0%	5.0%	10.0%	20.0%	26.7%	VV
-=	% of colum <u>n</u>	100.0	21.7	12.4	8.2	7.9	15.2	14.7	
:	% of total	0.2	4.6	0.7	0.7	1.5	2.9	3.9	14.5%
	Total Count	1	88	24	37	76	79	109	413
	% of Cases	0.2%	21.2%	5.9%	8.9%	18.3%	19.2%	26.3%	100.0%
									۸۳

As Table III-5 shows, the ADWs tended to be reported to the police promptly; 71.3% were reported within 1 hour after occurrence, and another 9.0% between 1 and 2 hours after occurrence. Few reports were made beyond one day (19 cases).

Approximately half the crimes occurred in buildings; the remainder took place in the street or in a park or recreational area (see Table III-6). In the case of crime location, we note a difference among the three clearance categories, with 53% of the cleared cases, 55.7% of the cleared-other cases and only 33.3% of the uncleared cases occurring in a building. A similar difference is shown in Table III-7, which gives the facility category where a crime took place. Although 47% of the cleared cases and 54.4% of the cleared-other cases took place in residences, only one-fourth of the uncleared cases occurred in residential facilities. Thus a significantly higher number of cleared than uncleared cases occurred inside, with the uncleared cases being predominantly street crimes.

Table III-8 shows the weapons used in the ADWs. The most common weapon was a handgun: 33.4% of the cases involved handguns. Another 23.4% involved knives, and 10.9% involved the use of a blunt instrument. In 24.3% of the cases the weapons used were classified as "other." This typically was bodily force, because an assault can be classified as an ADW when the suspect is sufficiently stronger than the victim to inflict on him severe bodily harm. Rifles, shotguns, and alleged guns accounted for only 8 of the 413 ADWs, with the weapon used either unknown or not indicated in 25 cases.

Almost four out of five ADWs were committed by a single offender (see Table III-9). This percentage is lower for the uncleared cases, where three out of five crimes involved one offender and another 21.7% involved two offenders. In only 28 out of the total sample of 413 cases were more than two offenders involved.



Table ILI-5

ADW: CLEARANCE BY TIME BETWEEN OCCURRENCE AND REPORT

	.Status	Unknown Time	Within 1 Hour	l to 2 Hours	2 to 4 Hours	4 to 8 Hours	8 to 12 Hours	12 to 24 Hours	l to 2 Days	2 to 4 Days	4 to 7 Days	7 to 10 <u>Days</u>	10 to 14 Days	l4 to 2l Days	<u>Total</u>
	Cleared														
	Count	1	153	13	8	6	Ĵ	10	Ţ	2]	()	Ď	İ	206
	% of row	0.7%	74.3%	6.5%	4.1%	3,1%	1.7%	5.1%	0.7%	1.2%	1,4%	0%	0.5%	0.7%	
	% of column	24.9	51.9	36.0	56.7	54,]	63.5	45.8	30.0	55.3	39.9	0	100.0	100.0	
	% of total	0.4	37.0	3, 2	2.0	1,6	0,8	2.5	0.4	0.6	0.7	0	0.2	0.4	49.9%
	Cleared-Other														
	Count	3	103	15	3	4	Q	11	j	Õ)	0	0	0	147
	% of row	2.3%	69.8%	10.1%	2.3%	3.0%	0%	7.7%	2.3%	0%	2, 3%	0%	0%	0%	
Ŋ	% of column	58.2	34.8	39.8	23, 2	37.5	0	49.8	70.0	0	46.6	0	Ō	0	
7	% of total	0.8	24.8	3.6	0.8	1.1	Õ	2.7	0.8	Õ	0.8	0	0	0	35.6%
	Uncleared														
	Count	Ţ	39	9	3	1	2	1	0	2	1	1	Q	Ö	60
	% of row	1.7%	65.0%	15.0%	5.0%	1.7%	3.3%	1.7%	0%	3.3%	1.7%	1.7%	0%	0%	**
	% of column	16.9	13.2	24.2	20.2	8.4	36.5	4.4	0	44,7	13,5	100.0	0	0	
	% of total	0.2	9.4	2.2	0.7	0,2	0.5	0.2		0.5	0.2	0.2		()	14.5%
	Total Count	6	295	37	15	12	<u>.</u>	23	5	4	7	1	1	l	413
	% of Cases	1,4%	71.3%	9.0%	3.6%	2.9%	1.3%	5,5%	1.2%	1.1%	<u>l</u> . 8%	0.2%	0.2%	0.4%	100.0%

Table III-6
CLEARANCE BY LOCATION OF CRIME

Park or Recreational Unknown Street Area Building Total Cleared Count 6 87 4 109 206 % of row 2.9% 42.2% 53.0% 1.9% % of column 100.0 45.5 79.8 51.7 % of total 1.4 21.1 1.0 26.4 49.9% Cleared-Other Count 0 65 0 82 147 % of row 0% 44.3% 0% 55.7% % of column . 0 34.1 0 38.8 % of total 0 15.8 0 -19.8 35.6% Uncleared Count 0 39 1 20 60 % of row 1.7% 0% 65.0% 33.3% % of column 0 20.4 20.2 9.5 % of total 0 9.4 0.2 4.8 14.5% Total Count 191 5 211 413

ADW:

1.4%

46.3%

1.2%

51.1%

100.0%

% of Cases



Table III-7

ADW: CLEARANCE BY FACILITY CATEGORY

Status	<u>Unknown</u>	Residential	<u>Commercial</u>	<u>Public</u>	Transpor- tation	<u>Total</u>
Cleared	i					
Count	75	97	16	7	11	557
% of row	36.5%	47.0%	7.7%	3.6%	5.3%	206
% of column	47.9	50,5	49.8	100.0	2.2% 43.1	
% of total	18.2	23,4	3.8	1.8	2.6	49.9%
Cleared-Other						
Count	47	80	9	0	11	147
% of row	31.9%	54.4%	6.1%	0%	7.7%	± ¬ (
% of column	29.8	41.7	28.1	0	45.0	
% of total	11.3	19.3	2.2	0	2.7	35.6%
Uncleared		·				
Count	35	15	7	Ó	3	/ñ
% of row	58.3%	25.0%	11.7%	0%	ر 5.0%	60
% of column	22.3	7.8	22.1	0	11.9	
% of total	8,5	3.6	1.7		0.7	14.5%
Total Count	157	192	3 <u>2</u>	7	25	413
% of Cases	38.0%	46.4%	7.7%	1.8%	6.1%	100.0%

100

Table III-8

ADW: CLEARANCE BY WEAPON USED

Status	No Weapon	Handgun	<u>Rifle</u>	<u>Shotgun</u>	Alleged Gun	<u>Knife</u>	Blunt Instrument	Other <u>Weapon</u>	Unknown Weapon	<u>Total</u>
Cleared										
Count	12	61	4	2	0	45	23	59	Ō	206
% of row	6.0%	29.4%	1.9%	1.2%	0%,	21.8%	11.1%	28.6%	0%	
% of column	54.1	44.0	100.0	71.2	0	46.6	50.6	58.6	0	
% of total	3.0	14.7	1,0	0.6	. 0	10.9	5.5	14.3	0	49.9%
Cleared-Other										
Count	. 3	61	0	0	0.	36	18	28	0	147
% of row	2.3%	41.6%	0%	0%	0%	24.8%	12.4%	18.8%	0%	
% of column	15.1	44.4	0	0	0	37.3	40.5	27.5	0	
% of total	0.8	14.8	Ō	0	0	8.8	4.4	6.7	0	35.6%
Uncleared										
Count	7	16	0	1	1	15	4	14	2	60
% of row	11.7%	26.7%	0%	1.7%	1.7%	25.0%	6.7%	23.3%	3.3%	
% of column	30.7	11.6	0	28.8	100.0	15.6	8.9	13.9	100.0	
% of total	1.7	<u>3.9</u>	0	0.2	0.2	3,6	1.0	3.4	0.5	14.5%
Total Count	23	138	4	3]	96	45	100	2	413
% of Cases	5.5%	33.4%	1.0%	0.8%	0.2%	23.4%	10.9%	24.3%	0.5%	100.0%

Table III-9

ADW: CLEARANCE BY NUMBER OF OFFENDERS

Status	Not Known	<u>One</u>	Two	Three	Pour	<u>Five</u>	Total
Cleared							
Count '	1	173	24	3	4	1	206
% of row	0.5%	84.1%	11.6%	1.4%	1.9%	0.5%	
% of column	33.3	52.9	44.0	18.1	44.2	33.3.	
% of total	0.2	42.0	5.8	0.7	1.0	0.2	49.9%
Cleared-Other							
Count	0	118	17	10	Ō	1	147
% of row	0%	80.5%	11.7%	7.0%	0%	0.7%	
% of column	0	36.1	31.9	63.5	0	33.3	
% of total	0	28.7	4.2	2.5	0	0.2	35.6%
N							ı
 Uncleared							
Count :	2	36	13	3	5	1	60
% of row	3.3%	60.0%	21.7%	5.0%	8.3%	1.7%	
% of column	66.7	11.0	24.0	18.4	55.8	33.3	
% of total	0.5	8.7	<u>3.1</u>	0.7	1.2	0.2	14.5%
Total Count	3	328	54	16	9	3	413
% of Cases	0.7 %	79.3%	13.1%	3.9%	2.2%	0.7%	100.0%

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Not only did ADWs tend to be reported promptly, but where an arrest was made, it was generally made within 1 hour of the time of report (82% of the arrests). Table III-10 gives the time between the report of the crime and the arrest of a suspect and shows whether a suspect's name had been given to the police at the time of report. (Suspect named is broken down into: real name given, also known as (AKA) given, partial name given, and nickname given.) Arrests were made in 55 cases where the suspect had not been named. However, 49 of these arrests occurred within 1 hour of the report of the offense. Clearly, ADW can be characterized as a crime generally committed by a person known to the victim. When an unnamed suspect was arrested, the arrest was generally within 1 hour of the report of the crime.

Bivariate correlations were run with 105 variables. The cleared and cleared-other cases formed one group, and the uncleared cases were another group. Thus the closer the bivariate correlation is to one, the more closely associated with clearance is the variable. Listed in Table III-11 are the variables showing at least a 0.15 correlation with clearance, in order of descending correlation (i.e., the first has the highest association with clearance).

Only two arrests were made after 8 hours from the time of report where a suspect had not been named (after 8 hours it is reasonable to assume that patrol's input had ended and an investigator had received control of the case). Consequently, it was decided that a follow-up investigation decision rule could not realistically be constructed for ADW.

Although a follow-up decision rule was not constructed, a discriminant analysis was run to illustrate the variables contributing to clearing ADWs. All variables with a correlation coefficient of at least 0.1 were included. Cases were excluded where an arrest was made less than 8 hours after report. Cases where an arrest was made more than 8 hours after report and



ADW: SUSPECT NAMED BY TIME BETWEEN REPORT AND ARREST

Status	No <u>Arrest</u>	Within One <u>Hour</u>	l to 2 Hours	2 to 4 Hours	4 to 8 Hours	12 to 24 Hours	2 to 4	4 to 7 Days	7 to 10 Days	10 to 14 Days
Not named										
Convi	60	49	3	1	0	1	1	٨	A	_
% of row	52.4%	42.4%	2,6%	0.9%	0%	0.9%	0.9%	0	, Õ	0
% of column	28.3	29.9	33.3	22.5	Ō	18.3	33.3	0% ^	0%	0%
% of total	14.6	11.8	0.7	0.2	0	0.2	0, <u>2</u>	Q Q	0 0	0
Real Name			ś	i			* • =	Ą	U	0
Count	140	109	6	, 3	1	i.	A			
% of row	50.5%	39.3%	2.1%	1.2%		4	2	1	2	1
% of column	65,9	66.8	66.7	77.5	0.5% 100.0	1.6%	0.7%	0.5%	0.7%	0.5%
% of total	34.0	26.4	1.4		100.0	81.7	66.7	100.0	100.0	100.0
AKA-Also Known As		20:7	± 4 4	0.8	0.4	1.1	0.5	0.4	0.5	0,4
Count .	ň	i	ā							
% of row	0 0%	100 00	Ō	0	0	0	0	0	. 0	Ō
% of column	•	100,0%	0%	0%	0%	0%	0%	0%	0%	0%
% of total	Ò	0.9	0	0	0	Ô	Ō	Ô	Q	0
% OI COCHI	0	0.4	0	0	0	0	Ō	0	٥	0
Partial Name										Ů
Count	8	l	0	Ò	0	Q.	Ō	0	٥	
% of row	66.7%	17,4%	٥%	0%	0%	0%	0%	0%	Ō	0
% of column	3.7	0.9	0	0	0	0	0/2		0%	0%
% of total	1,9	0.4	0	0	0	0	0	0	0	0
Nickname				-	ų	V	V	0	0	. 0
Count	4	2	0	Õ	Α	_				
% of row	64.4%	35.6%			()	0	0	0	0	0
% of column	2.1	1.5	0%	0%	0%	0%	0%	0%	0%	Ō%
% of total			0	0	0	0	0	0	0	Ô
	1.1	0.6	0	0	0	0	0	0	<u> </u>	0
Total Count	213	163	9	4	1	5	3	1		1
% of Cases	51.4%	19.6%	2.1%	1.1%	0.4%	1.3%	0.7%	0.4%	0.5%	0.4%

W

Table III-11

ADW VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS

Variable	Correlation Coefficient
	2110
Suspect known	J . 5112
Suspect named at time of report	0.4948
Words spoken by offender	0.4279
Suspect previously seen	0.2870
Offender and victim of same race	0.2547
Black offender/black victim	0.2533
Weapons as evidence	0.2477
Victim invited offender in	0.2341
Black victim	0.2306
Suspect's associates named or indicated	0.2297
Places suspect frequented named	0.2081
Weapon match	0.1996
Offender violent	0.1963
One offender	0.1809
Crime location building	0.1540

cleared and cleared-other cases where no arrest was made were considered as one group; the uncleared cases formed the other group. The eight variables that exhibited the largest discriminant function coefficients are:

- Suspect named (more than twice the size of the next coefficient)
- License number given
- Suspect known
- Black victim/black offender
- Words spoken by offender



- Weapons as evidence and match
- Victim invited offender in
- Suspect previously seen.

These variables, although not suitable for the construction of a decision rule, nevertheless indicate the type of information most likely to contribute to ADW case clearances.



CHAPTER IV. CAR THEFT

Motor vehicle theft was the highest-volume crime analyzed. In the three-month sample period, there were 1187 motor vehicle thefts (California Vehicle Code 10851). The crime also showed the lowest clearance rate of any of the crimes investigated, with 8.8% classified cleared and another 3.2% classified as cleared-other (see Table I-8).

The reasons for this low clearance rate are clear. In the other crimes coded, there is at least a brief offender/victim confrontation. In car theft cases, the victim generally has no idea who stole his vehicle nor, in many cases, does he know the time when it was stolen. This leaves the police investigator with very little information on which to base his investigation.

In response to these issues, the OPD procedure for handling cases of motor vehicle theft is different from that for other felony crimes. The report is taken over the telephone by a police technician rather than by a patrol officer at the scene, unless the crime is "in progress." Then efforts are concentrated on recovering the vehicle rather than on apprehending an offender. If the vehicle is occupied when it is recovered, the occupants are obviously booked for motor vehicle theft.

Because of this procedure, the construction of a case follow-up decision rule for motor vehicle theft was not technically feasible. In this chapter, however, we present a number of interesting cross tabulations and correlations gleaned from our data.

Unlike the ADW cases, where over 71% of the offenses were reported to the police within 1 hour (see Chapter III), only 12.3% of the car thefts were reported within an hour of occurrence (see Table IV-1).

Table IV-1

CAR THEFT: CLEARANCE BY TIME BETWEEN OCCURRENCE AND REPORT

Status	Unknown Time	Within 1 Hour	1 to 2 Hours	2 to 4 Hours	4 to 8 Hours	8 to 12 Hours	12 to 24 Hours	1 to 2 Days	2 to 4 <u>Days</u>	4 to 7 Days	30 to 45 <u>Days</u>	Total
Cleared									·			
Çount	21	32	9	10	10	8	6	5	1	1	1	104
% of row	20.2%	30.8%	8.7%	9.6%	9.6%	7.7%	5,8%	4.8%	1,0%	1.0%	1.0%	
% of column	27.4	21,9	7,4	6,4	6.0	3,4	3.8	8,8	2,6	3,4	50,0	
% of total	1.8	2,7	0,8	0,8	0,8	0.7	0,5	0.4	0.1	0,1	0.1	8.8%
Cleared-Other						;						
Count	7	3	4	3	6	3	2	4	4	1	1,	38
% of row	18,4%	7.9%	10,5%	7.9%	15.8%	7.9%	5,3%	10,5%	10,5%	2,6%	2,6%	
% of column	9,1	2,1	3,3	1,9	3.6	1.3	1,3	7.1	10,2	3,4	50.0	
% of total	0.6	0.3	0,3	0,3	0,5	0.3	0,2	0,3	0.3	0,1	0.1	3.2%
Uncleared	• •											
Count	49	111	109	143	150	225	150	48	34	27	0	1,045
% of row	4.7%	10.6%	10,4%	13.7%	14,3%	21,5%	14,3%	4,6%	3,3%	2,6%	0%	
% of column	63,5	76.0	89.3	91.7	90,4	95.3	94.9	84.1	87.2	93.2	0	
% of total	4,1	9,3	9,2	12.0	12.6	18.9	12.6	4,0	<u>2.9</u>	2.3		88.0%
Total Count	77	146	122	156	166	236	158	57	39	29	2	1,187
% of Cases	6.5%	12,3%	10.3%	13.1%	14.0%	19.9%	13,3%	4.8%	3.3%	2,5%	0.2%	100.0%

However, nearly 83% of the car theft cases were reported within the first 24 hours after occurrence.

As has been noted, the time of occurrence of the car theft is also often in doubt (see Table IV-2). No information regarding time of occurrence was given in 3.4% of the cases. In another 72.4% of the cases, the time of occurrence was given as a range of time, e.g., between 1600 and 2400 hours. In only 24.2% of the cases was the victim or a witness able to state exactly when the theft took place. However, the time of occurrence was certain in 37.5% of the cleared cases and in 39.5% of the cleared-other cases.

Table IV-3 is presented to illustrate the paucity of information available in the cases. In 84.5% of the cases, the sex of the offender was not known at the time of report. However, this fact was unknown in only 25% of the cleared cases and 10.5% of the cleared-other cases, which indicates that the ability to describe a suspect contributes to clearing the case.

A lower percentage of the suspects in the car theft cases had previously been seen, known, or named than in the other felonies investigated (see Tables IV-4, -5, and -6).* Nevertheless, these variables were important contributors to clearance, with the suspect having previously been seen in 22.1% of the cleared cases and 55.3% of the cleared-other cases. The suspect was known in 19.2% of the cleared and 63.1% of the cleared-other cases and was named in 21.2% of the cleared and 65.9% of the cleared-other cases.

Ψ.

The 20 uncleared cases where a suspect was named might seem puzzling. Actually, they are 3 cases in our coding (weighted to 20) where a person gave a name, presumably his own, when renting a car which he subsequently neglected to return. The cars were recovered, but the suspects were no longer in them.

Table IV-2

CAR THEFT: CLEARANCE BY CERTAINTY OF TIME OF OCCURRENCE

Status	No Time	Certain Time	Uncertain Time	<u>Total</u>
Cleared				
Count	14	39	51	104
% of row	13.5%	37.5%	49.0%	
% of column	34.6	13.6	5.9	
% of total	1.2	3.3	4.3	8.8%
Cleared-Other				
Count	6	15	17	38
% of row	15.8%	39.5%	44.7%	
% of column	14.8	5.2	2.0	
% of total	0.5	1.3	1.4	3.2%
Uncleared				*
Count	20	234	791	1,045
% of row	2.0%	22.3%	75.7%	•
% of column	50.5	81.2	92.1	
% of total	1.7	19.7	<u>66.6</u>	88.0%
Total Count	40	288	859	1,187
% of cases	3.4%	24.2%	72.4%	100.0%

Table IV-3

CAR THEFT: CLEARANCE BY OFFENDER SEX

Status	Not Known	<u>Female</u>	<u>Male</u>	Total
Cleared				
Count	26	4	74	104
% of row	25.0%	3.8%	72.1%	
% of column	2.8	12.7	31.7	
% of total	2.2	0.3	6.2	8.8%
Cleared-Other	or.			
Count	4	6	28	38
% of row	10.5%	15.8%	73.7%	
% of column	0.4	19.1	12.0	
% of total	0,3	0.5	2.4	3.2%
Uncleared	. •			
Count	892	21	131	1,045
% of row	85.4%	2.1%	12.6	•
% of column	96.7	68.2	56.3	
% of total	75.2	1.8	11.1	88.0%
Total Count	922	31	233	1,187
% of Cases'	77.7%	2.6%	19.7%	100.0%



Table IV-4

CAR THEFT: CLEARANCE BY SUSPECT PREVIOUSLY SEEN

landa da karaka karaka da	Not Pre-			Seen by		
	viously	Seen by	Seen by	Citizen	Seen by	
Status	Seen	Victim	<u>Witness</u>	Informant	Police	Tota1
Cleared	x .	,				·
Count	81	. 15	7	0	1	104
% of row	77.9%	14.4%	6.7%	" 0%	1.0%	
% of column	7.4	24.5	29.6	0 .	100.0	
% of total	6.8	1.3	0.6	0	0.1	8.8%
Cleared-Other						
Count	17	19	2	0	0	38
% of row	44.7%	50.0%	5.3%	0%	0%	
% of column	1.6	31.0	8.5	0	o	
% of total	1.4	1.6	0.2	0	o	3.2%
Uncleared						
Count	996	27	.15	7	0	1,045
% of row	95.3%	2.6%	1.4%	0.7%	0%	•
% of column	91.0	44.5	61.9	100.0	O	
% of total	83.9	2.3	1.2	0.6	0,	88.0%
Total Count	1,094	61	24	· 7	1	1,187
% of Cases	92.2%	5.2%	2,0%	0.6%	0.1%	100.0%

Table IV-5
CAR THEFT: CLEARANCE BY SUSPECT KNOWN

16. 1						
			Known	Known	Known	
	*	Not	to	to	to	
	Status	Known	<u>Victim</u>	<u>Witness</u>	<u>Police</u>	<u>Total</u>

100	Cleared			_	_	
**	Count	84	17	2	.1	104
	% of row	80.8%	16.3%	1.9%	1.0%	
	% of column	7.4	36.3	18.5	100.0	
	% of total	7.1	1.4	0.2	0, 1	8.8%
	Cleared-Other					
No.	Count	14	23	1	. 0	3 8
	% of row	36.8%	60.5%	2.6%	0%	•
	% of column	1.2	49.1	9.3	0	
· ·	% of total	1.2	1.9	0.1	0	3.2%
	Uncleared	•				
7	Count	1,030	7	8	0	1,045
* .5	% of row	98.6%	0.7%	0.7%	0%	
	% of column	91.3	14.5	72.2	0	
	% of total	86.8	0.6	0.7	0	88. 0%
	Total Count	1,128	47	11	1	1,187
	Total count	1,120	- * *			v -
	% of Cases	95.1%	3.9%	0.9%	0.1%	100.0%



And the second

Table IV-6

CAR THEFT: CLEARANCE BY SUSPECT NAMED

	Status	Not Named	Real Name	AKA-Also Known As	Partial Name	Nickname	Total
	Cleared						
# : · *	Count	82	20	0	2	5 0	104
i de la serie. Maria	% of row	78.8%	19.2%	0%	1.9%	0%	
	% of column	7.3	32.6	0	50.0	0	
	% of total	6.9	1.7	0	0.2	O	8.8%
	Cleared-Other					• • • ,	
	Count	13	21	2	2	0	38
	% of row	34.2%	55.3%	5.3%	5.3%	0%	
	% of column	1.2	34.2	100.0	50.0	0	
1:	% of total	1.1	1.8	0.2	0.2	O	3.2%
	Uncleared						
ž.	Count	1,024	20	0	0	1	1,045
	% of row	97.9%	2.0%	0%	0%	0.1%	•
	% of column	91.5	33.3	0	0	100.0	
	% of total	86.2	1.7	0	0	0.1	88.0%
	Total Count	1,119	61	2	4	1 .	1,187
	% of Cases	94.2%	5.2%	0.2%	0.3%	0.1%	100.0%

e in many expression

Because of the procedures followed by the OPD--an arrest is made if the car is occupied when it is recovered--the time between report and arrest is more scattered than for the other felonies investigated (Table IV-7). Of the 107 arrests, 33 occurred within 1 hour of the time of report of the theft; but another 21 occurred 1 to 2 days after the time of report, and 11 occurred within 2 to 4 days after the time of report. Two arrests occurred in each of the following categories: 10 to 14 days after report; 14 to 21 days after report; and 21 to 30 days after report. One arrest occurred more than 45 days after report.

As mentioned above, the priority of the OPD in dealing with car theft cases is to recover the car (see Table IV-8). During the sampling period, 94.4% of the vehicles were recovered. This percentage includes 101 recoveries where cases were cleared, 37 recoveries where cases were cleared-other, and 983 recoveries where there were no clearances.

Despite our belief that it was not possible to construct a follow-up decision rule, we did run both bivariate correlations and a discriminant analysis with the motor vehicle theft cases to determine the relative importance of each variable in contributing to clearance of the cases.

The "8 hours after time of report rule" was not followed in the car theft cases, because patrol typically is not involved in these cases, which are usually reported by telephone. Thus the sample was considered as a whole. The variables in Table IV-9 had a correlation coefficient of at least 0.15 with clearance (cleared and cleared-other combined). Only the variables where we had data in most of the cases were included (i.e., where there was not an overwhelming number of missing values, as was often the case). Again, they are listed in order of descending correlation with clearance.

A discriminant analysis was run with these variables, as well as with several others having even weaker correlation with clearance. (A cutoff of 0.1 was used.) The nine variables found to have the largest



Table IV-8

CAR THEFT: CLEARANCE BY CAR RECOVERED

Not Status Recovered Recovered Total Cleared 3 104 Count 101 % of row 2.9% 97.1% % of column 9.0 4.5 % of total 8.5 8.8% 0.3 Cleared-Other Count 1 37 38 % of row 2.6% 97.4% % of column 1.5 3.3 % of total 3.2% 0.1 3.1 Uncleared 1,045 Count 62 983 % of row 6.0% 94.0% % of column 94.0 87.7 88.0% % of total 82.8 5,2 1,121 Total Count 66 1,187 94.4% 100.0% % of Cases 5.6%

Table IV-9

CAR THEFT VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS

	Correlation
Variable	Coefficient
Suspect description developed (positive	
if race, sex, or age given)	0.5070
Suspect known	0,4431
Suspect named	0.4323
Suspect previously seen	0.3185
On-view report of crime	0.2964
Suspect associates named/indicated	0.2741
Vehicle registration check, useful lead	0.2738
Offender invited suspect in (typically	
offender took advantage of owner)	0.2728
Facility category-residential	0.2584
Places suspect frequented named	0.2277
Fingerprint match	0.1934
Direction of flight provided	0.1944
Crime lab report	0.1764
Time between occurrence and report less	
than one hour	0.1757
Time of occurrence certain	0.1527

124

10 pt 2 m 24

discriminant function coefficients are: suspect description developed; vehicle registration check useful; on-view report of offense; suspect named; time of occurrence between 0400 and 0800 hours; suspect known; time between occurrence and report (a negative coefficient indicating that the longer the time, the lower the probability of clearance); other physical evidence present; and victim invited offender in.

These variables, while not suitable for the construction of a decision rule because of the lack of follow-up investigation in the Department, nevertheless indicate that the ability to develop any information regarding a suspect is the key to solving a car theft case. In the absence of such information, apprehension is largely a random event.



CHAPTER V. RAPE

Rape was the lowest-volume crime category that we analyzed. In the three-month sample there were a total of 65 reported cases; all were coded for computer processing. Rape was chosen as a felony for analysis for various reasons:

- It is an FBI Part I crime against person.
- · It is a traumatic experience for the victim.
- It is one of the most difficult crimes to prosecute.

The crime clearance classification for our analysis is shown in Table V-1. It can be seen that approximately 21% of the forcible rape cases were cleared. The same percentage of the forcible rape cases were classified cleared-other under our criteria. The total clearance rate for forcible rape was 42.8%. For attempted rape, approximately 30% of the cases were classified cleared, and 13% were cleared-other. The total clearance rate for attempted rape was 43.4%.

Differences between our classification and the case disposition taken by the OPD have been shown in Table I-6. This table has shown that we classified nearly 25% of the reported cases as cleared and nearly 19% of the reported cases as cleared-other. This totals to about 43% overall clearance. On the other hand, by using the OPD classification procedure, 60% overall clearance would be shown.

The rape cases differed in many respects from the other felony cases we studied. Our analysis of other person-to-person crimes (robbery and ADW), showed that the suspects named and being known dominated the other indicators contributing to case solution. This was not true for the rape cases. It can be seen from Table V-2 that 37.8% of the uncleared cases



Table V-1

RAPE: CLEARANCE BY PRIMARY FELONY OFFENSE

Status	Forcible Rape	Attempted Rape	<u>Total</u>
Cleared			
Count	9	7	16
% of row	56.3%	43.8%	
$% \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \$	21.4	30.4	
% of total	13.8	10.8	24.6%
Cleared-Other			
Count	9	3	12
% of row	75.0%	25.0%	
% of column	21.4	13.0	
% of total	13.8	4.6	18.5%
Uncleared	· · · · · · · · · · · · · · · · · · ·		
Count	24	13	37
% of row	64.9%	35.1%	
% of column	57.1	56.5	
% of total	<u>36.9</u>	20.0	56.9%
Total Count	42	23	65
% of Cases	64.6%	35.4%	100.0%

Table V-2

RAPE: CLEARANCE BY SUSPECT NAMED

Status	Not Named	Real <u>Name</u>	Partial Name	Nickname	Total
Cleared					
Count	8	7	1	0	16
% of row	50.0%	43.8%	6.3%	0%	
% of column	22.9	41.2	10.0	0	
% of total	12.3	10.8	1.5		24.6%
Cleared-Other					1
Count	4	5	1	2	12
% of row	33.3%	41.7%	8.3%	16.7%	
% of column	11.4	29.4	10.0	66.7	
% of total	6.2	7.7	1.5	3.1	18.5%
Uncleared					
Count	23	5	8	1	37
% of row	62.2%	13.5%	21.6%	2.7%	
% of column	65.7	29.4	80.0	33. 3 ,	
% of total	<u>35.4</u>	7.7	12.3	1.5	56.9%
Total Count	35	17	10	3	65
% of Cases	53.8%	26.2%	15.4%	4.6%	100.0%

remained uncleared even though the suspects had been named. (In ADW, for example, only 18.3% of the uncleared cases showed named suspects.) Similar differences appeared when the suspect-known variable was analyzed. It seemed that this element of information was not being (or could not be) used as effectively in the rape cases as in other person-to-person crimes. On the other hand, rape cannot be characterized as a stranger-to-stranger type of crime. Table V-2 shows that, in 46.2% of all cases, some name was present. Additionally, in 35.4% of the cases, the offender was known by someone at the scene, and in 24.6% of the cases the offender had been seen previously.



The remainder of t's chapter records some information gleaned from our preliminary rape analysis using cross tabulations, followed by some of the results of the correlation analysis. Although it was not feasible to construct a decision rule for follow-up investigation of rape cases, we performed a discriminant analysis of screened variables to try to identify the information elements that contributed most to case clearance.

One aspect that we analyzed initially was the relationship of the elapsed time between report and arrest to clearance (Table V-3 shows the cross tabulation). In 8 of the 16 cases in the cleared category, arrests

Table V-3

RAPE: CLEARANCE BY TIME BETWEEN REPORT AND ARREST

Status	No Arrest	Within 1 Hour	1 to 2 Hours	Total
Cleared				
Count	8	6	2	16
% of row	50.0%	37.5%	12.5%	
$% \ \ \text{of column}$	14.0	100.0	100.0	
% of total	12.3	9.2	3.1	24.6%
Cleared-Other				
Count	12	0	0	12
% of row	100.0%	0%	0%	
% of column	21.1	0	0	
% of total	18.5	. 0	0	18.5%
Uncleared				
Count	37	0	0	37
% of row	100.0%	0%	0%	
% of column	64.9	0	0	
% of total	56.9	0	0	56.9%
Total Count	57	6	2	65
% of Cases	87.7%	9.2%	3.1%	100.0%



were made within the first 2 hours after report. Half the reports of the cleared cases did not indicate the lapse of time between report and arrest. Of these 8 cases, 2 involved nonarrest types of legal action (D.A. Citation, Notice To Appear); 2 were classified as cleared on the basis of warrants having been issued. The remaining cases had notations such as the suspect being "out on week-end release," or no arrest was indicated because the suspect was already being held for another offense. Since most of the arrests occurred within 2 hours, it must be assumed that patrol is making the most significant contribution to clearance.

We next looked at the relationship of the elapsed time between occurrence and report to case clearance. (Table V-4 shows the cross tabulation.) In 14 of the 16 cleared cases, the incidents were reported within 12 hours of occurrence. Ten out of 12 cleared-other cases were reported before 12 hours. Less than half (39%) of all the clearances were effected when the time between incident and report did not exceed 1 hour. For about half (32 cases) of the 65 total cases sampled, time lapses of less than 1 hour were indicated. But only 28% of these were cleared (9 cases), and 6% were classified as cleared-other (2 cases). There seems to be some indication of a higher clearance rate when incidents were reported quickly; but the effect was not dramatic, because about half the total cases (including the uncleared cases) were reported quickly.

The victims in the cleared-other category seem to have shown a basic reluctance to report the crimes. Note that only 16.7% of these cases were reported in the first hour; in the other categories the figure was about 50%. In most of the cases in the cleared-other category, the victims named the offenders in the initial reports and then failed to respond to the OPD investigators' attempts to contact them. The OPD often clears such cases (as well as cases where no offender was named) as "Complainant Refuses To Prosecute."



RAPE: CLEARANCE BY TIME BETWEEN OCCURRENCE AND REPORT

Table V-4

	Status	Unknown Time	Within 1 Hour	l to 2 Hours	2 to 4 Hours	4 to 8 Hours	8 to 12 Hours	12 to 24 Hours	4 to 7 Days	Total	
	Cleared										
	Count	1	9	3	1	0	1	1	Ŏ	16	
	% of row	6.3%	56.3%	18.8%	6.3%	0%	6.3%	6.3%	0%	Ŧħ	
as kan tir	% of column	50.0	28.1	25.0	20.0	0	33.3	25.Ū) ()		
:	% of total	1.5	13.8	4,6	1.5	0	1.5	1.5	0	24.6%	
	Cleared-Other					i					
Q) Q)	Count	1	2	3	2	2	1	1	0	12	
<u>a</u> ,	% of row	8.3%	16.7%	25.0%	16.7%	- 16.7%	8.3 %	8.3%	0%	Ŧ <i>₹</i>	
	% of column	50.0	6.3	25.0	40.0	33.3	33.3	25.0	0		
	% of total	1.5	3.1	4.6	3.1	3.1	1.5	1.5	0	18.5%	
÷	Uncleared			ı							
	Count	0	21	6	2	4	1	2	1	רֿיִּ	
	% of row	0%	56.8%	16.2%	5.4%	10.8%	2.7%	5.4%	2.7%	37	
	% of column	0	65.6	50.0	40.0	66.7	33.3	50.0	100.0		
	% of total	0	<u>32.3</u>	9.2	3.1	6.2	1.5	3.1	1.5	56.9%	
n i	Total Count	2	32	12	5	6	3	4	1	65	132
31	% of Cases	3.1%	49.2%	18.5%	7.7%	9.2%	4.6%	6.2%	1.5%	100.0%	T () (i

The next aspect we considered was the location of the crimes. We noticed that frequently the victim had been moved from the location of initial contact. Most of the rape cases were reported as having occurred with a street contact of some kind. Figure V-1 is a bar graph of the two primary locations: street and building. From this figure it can be seen that a crime that had no aspect of street contact was four times as likely to be in the cleared category as one that did.

After the initial contact, most rape offenders and victims moved to a place of relative privacy (if the initial contact was not in a private location). The facility category is the best indicator of where the crimes actually took place. The cross tabulation of this variable is shown in Table V-5. It can be seen that at least seven of the street victims were moved to residential facilities. Twenty-one of the street contacts apparently continued in transportation (all automobiles in this case). One must not draw the conclusion that, since a higher percentage of crimes

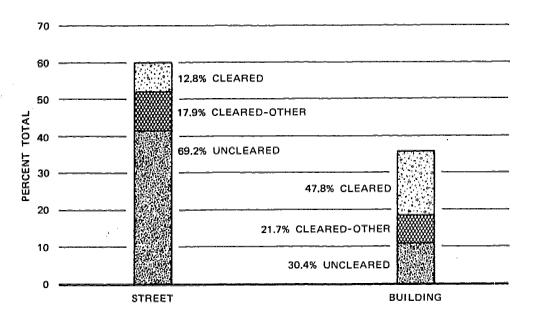


FIGURE V-1 CRIME LOCATION BY CLEARANCE CATEGORY



Table V-5

RAPE: CLEARANCE BY FACILITY CATEGORY

Status	<u>Unknown</u>	Residen- tial	Commer- cial	<u>Public</u>	Transpor tation	- Total
Cleared						
Count	5	8	1	0	2	16
% of row	31.3%	50.0%	6.3%	0%	12.5%	
% of column	41.7	26.7	100.0	0	9.5	•
% of total	7.7	12.3	1.5	0.	3.1	24.6%
Cleared-Other						
Count	1	7	0	0	4	12
% of row	8.3%	58.3%	0%	0%	33.3%	1.4
% of column	8.3	23.3	0	0	19.0	
% of total	1.5	10.8	0	. 0	6.2	18.5%
Uncleared						
Count	6	· 15	0	1	15	37
% of row	16.2%	40.5%	0%	2.7%	40.5%	3,
% of column	50.0	50.0		100.0	71.4	
% of total	9.2	23.1	0	1.5	23.1	56.9%
Total Count	12	30	1	1	21	65
% of Cases	18.5%	46.2%	1.5%	1.5%	32.3%	100.0%

are associated with residential facilities were cleared, this variable by itself is important. It must be considered along with the building/ street variable discussed previously.

We next abunsider the victim and witness descriptions and offenderrelated information as recorded in our data base. The items selected are those that could best be discussed in terms of our data base. There were many other aspects surrounding the crime that we elected not to code.

The data show that 37.2% of the victims who reported rape to the OPD were white, and 56.3% were black. There were 10 crimes committed against juvenile victims. The other 54 crimes were committed against



adults, and one victim's age was unknown. We noticed that a large number of the white victims and a lesser (but still significant) number of the black victims were described by the police as "known prostitutes." Such statements may have had a bearing on case clearance.

The offender in most rape cases is described as a black adult; 75% of the cases were committed by blacks. The physical characteristics describing offenders were quite varied.

Tables V-6, -7, and -8 show the offender/victim race relationship.

Black offenders assaulted white victims in 38% of all cases, and black offenders assaulted black victims in 35%. Other offender/victim race involvement was statistically minor by comparison.

Of the 28 cases cited as cleared (16) and cleared-other (12), the majority (57% or 16 of 28) were black offender/black victim cases. This statistic is to be compared to the 25% clearance level of black offender/white victim (7 cases out of 28). When both the offender and the victim were black, a higher percentage of cases were cleared.

Correlation coefficients for the rape case variables were derived by the procedure discussed in Appendix D.

The bivariate correlation coefficients of the elements of information contributing to case clearance are shown in Table V-9. Only the coefficients with a significance level of at least 0.125 are included.

The race variables showed high correlations with clearance. These variables and suspect named, black victim, suspect known, and suspect previously seen are probably all victim-supplied information. We believe that many of the offender/victim same-race situations occurred in cases where the offender was named and known.

Table V-6

RAPE: OFFENDER'S RACE BY VICTIM'S RACE--CLEARED CASES

		Victim				
	Offender	White	<u>Black</u>	<u>Mexican</u>	<u>Total</u>	
**************************************	··· "··White - ··· ·	and the second of the second	141 ang menangan an agam ng			T 00 2 1:00:014 78:0
	Count	0	1	0	1	
	% of row	0%	100.0%	0%		
ili. V	% of column	0	11.1	0		
	% of total	0	6.3	• • • • • • • • • • • • • • • • • • • •	6.3%	
	•					
e General	Black					
The state of the s	Count	5	9	0	13	
	% of row	38.5%	61.5%	0%		
	% of column	83.3	88.9	С		
5 + 1 + 5	% of total	31.3	50.0	0	.1.3%	
· · ·	Mexican					
f., 9	Count	1	0	1	2	
÷.	% of row	50.0%	0%	50.0%	-	х,
ie.	% of column	16.7	0	100.0		
	% of total	<u>6.3</u>	0	6.3	12.5%	
	Total Count	6	9	1	16	
Service of the servic	% of Cases	37.5%	56.3%	6.3%	100.0%	

*

egy Art (Art Line)

Table V-7

RAPE: OFFENDER'S RACE'BY VICTIM'S RACE--CLEARED-OTHER CASES

		Victim		
Offender	White	<u>Black</u>	Mexican	<u>Total</u>
White				
Count	1	0	0	1
% of row	100.0%	0%	0%	
% of column	33.3	0	0	
% of total	8.3	0	0	8.3%
Black .				
Count	2	8	0	10
'% of row	20.0%	80.0%	0%	
% of column	66.7	100.0	0	
% of total	16.7	66.7	0	83.3%
Mexican		•		
Count	o	0	1	1
% of row	0%	0%	100.0%	
% of column	0	Ó	100.0	
% of total	0	0	8.3	8.3%
Total Count	3	8	1	12
% of Cases	25.0%	66.7%	8.3%	100.0%

Table V-8

RAPE: OFFENDER'S RACE BY VICTIM'S RACE--UNCLEARED CASES

Victim Not American Offender Known White Black Mexican Indian Chinese Total Not known Count 0 1 0 0 0 0 1 % of row 0% 100.0% 0% 0% 0% 0% % of column 0 4.2 0 Ō 0 0 2.7% % of total 0 2.7 Ō 0 0 0 White 2 0 1 4 Count 0 1 0 % of row 0% 25.0% 50.0% 0% 0% 25.0% % of column 0 4.2 22.2 Ó 0 100.0 % of total 0 2.7 5.4 Ō Ō 2.7 10.8% Black 1 18 7 0 0 0 26 Count % of row 3.8% 69.2% 26.9% 0% 0% 0% 77.8 % of column 100.0 75.0 Ō 0 Ō 18.9 % of total 2.7 48.6 0 Ō 0 70.3% Mexican 0 3 0 1 Ó 5 Count 1 Ō% Ō% 20.0% 20.0% 0% % of row 60.0% % of column 100.0 100.0 0 12.5 0 Õ % of total 0 8.1 2.7 2.7 0 Ó 13.5% Other 0 Count Ō 1 0 0 1 0% 100.0% 0% 0% 0% 0% % of row 4.2 % of column 0 0 O 0 0 0 % of total 2.7 0 Ó 0 0 2.7% 9 1 Total Count 1 24 1 1 37 % of Cases 2.7% 64.9% 24.3% 2.7% 2.7% 2.7% 100.0%



Table V-9

RAPE VARIABLES DERIVED FROM BIVARIATE CORRELATION ANALYSIS

Variable	Correlation Coefficient
Suspect and victim same race	0.4067
Suspect named	0.3814
Black victim/black suspect	0.3714
Crime location-building	0.3580
Black victim	0.3335
Suspect known	0.3067
Physical force used and injury inflicted	0.2648
Description of physical attack mode	0.2374
Facility category-residential	0.2290
Suspect previously seen	0.2056
Clothing as evidence	0.2055
Suspect associates named	0.2055
Clothing match	0.1983
Weapon(s) as evidence	0.1985
Weapon match	0.1393

The importance of the crime location-building variable has already been discussed. The fact that physical force and injury to the victim were associated with the cleared cases is a measure of the seriousness of the crimes and aided in establishing a believable case (in court terms).

Although we had decided not to construct a follow-up decision rule for rape, we conducted two separate analyses to reveal several characteristics of the rape cases. We first performed a factor analysis on the



cases. Factor analysis is an analytical technique that can be used to reduce the number of variables under consideration in an analysis by establishing underlying relationships among them. This process then enables the variables to be rearranged or reduced in number. Factor analysis transforms a set of variables into a particular linear combination of variables that accounts for more of the variance in the data than any other linear combination of variables.

Our goal in the factor analysis was to establish factors that explained the characteristics of the cleared cases. All variables from our correlation coefficient runs exhibiting correlation coefficients of at least 0.100 with a significance level of at least 0.125 were included in the analysis. Only the cleared and cleared-other cases were considered because the objective was to determine the factors that these cases exhibited. The five dominating factors resulting from this analysis were the following combinations of variables:

- · Crime location-building and facility category-residential.
- Physical force used, injury inflicted, and description provided of physical attack mode.
- · Suspect named and suspect known.
- · Black victim/black offender and offender and victim same race.
- · Clothing as evidence and clothing match.

Second, a discriminant analysis was performed on the data. Two groups were used in the analysis: the cleared and cleared-other cases and the uncleared cases. All variables with a significance of at least 0.125 and a correlation coefficient of at least 0.100 were included. These seven variables exhibited the highest discriminant function coefficients:

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- Crime location-building. Cases occurring inside were more likely to be cleared than street cases.
- Condition of victim. Cases were more likely to be cleared if the crimes had resulted in injuries to the victims.
- Juvenile offender. Cases with juvenile offenders were more likely to be solved than cases with adult offenders.
- Black victim. Cases with a black victim were more likely to be solved than cases with a white victim.
- Clothing as evidence and clothing match. Clothing was an important factor in case solution.
- Offender and victim of the same race. These cases were solved at a higher rate than cases where offender and victim were of different races.
- Suspect named. Obviously this was an important factor in clearing a case.

Associated and separate section

CHAPTER VI. ANALYSIS OF OFFENDER CRIMINAL HISTORIES

A. Introduction

This chapter analyzes criminal histories of the suspects identified in our sample. The past criminal offenses were classified into 17 categories:

Strong-arm robbery

Theft, from person

Armed robbery

Theft, purse snatching

Felony assault

Theft, shoplifting

Burglary

Theft, other

Car theft

Narcotics and drugs

Homicide, willful

Stolen property

Forcible rape

Vehicle law violation

Attempted rape

Other

Not indicated

The offenders were classified into four groups on the basis of their most recent offense (ADW, car theft, robbery, and rape).

The histories were obtained from various OPD divisions.* Juvenile records are available only for offenses committed in Oakland and are usually destroyed when the person reaches 18. Therefore, the juvenile histories of adult offenders are incomplete. Analyses were made on the following characteristics of the offenders:

- · Race and sex
- Number of prior offenses

^{*}Offenders for whom no records could be found are excluded.





j.

- Type of prior offenses
- Time from first to most recent offense
- · Average number of prior offenses per year.

B. Summary

The major findings of the analyses are summarized below.

- Over 80% of the offenders were black. (The population of Oakland is approximately 43% black.)
- · Over 90% of the offenders were male.
- Of the felony crimes analyzed, the one in which the highest percentage of females participated was ADW (about 20% of the ADW offenders were females, compared to about 10% female participation in robbery and car theft).
- Over 80% of the offenders had prior offenses on record. The persons whose most recent offense was car theft or rape had a higher prior offense record (about 86%) than did those whose most recent offense was robbery or ADW (about 81%).
- Repeat offenders averaged more than seven prior offenses.
- The patterns of prior offenses varied somewhat according to the most-recent-offense grouping.
- The average age at first offense was about 3 years younger for the persons with the most recent offense of car theft or robbery (15.5 years) than for the ADW and rape offenders (18.5 years).
- On the average, the repeat offenders had criminal records covering 7.4 to 12.1 years. The 7.4-year criminal record average was that for the car theft offenders.
- The persons whose most recent offense was car theft showed the highest average number of offenses per year in crime. Their average was 1.8 offenses per year; persons in the other three classifications averaged 1.2 offenses per year.





1. Race and sex characteristics of offenders. Table VI-1 shows the distribution of offenders by race and most recent offense. This table also illustrates the sample size available for subsequent analyses. Note that there were only 16 offenders in the rape classification while there were over 130 in each of the other three classifications. Consequently, the characteristics of these 16 rape offenders may not be fully representative of Oakland rape offenders.

Table VI-1

DISTRIBUTION OF NUMBER OF OFFENDERS
BY RACE AND MOST RECENT OFFENSE

	Most Recent Offense				
Race	ADW	Auto theft	Robbery	Rape	
White	20	17	15	2	
Black	150	110	111	1.1	
Mexican	4	5	5	3	
American Indian	4	3	0	0	
Japanese	1	О	0	. 0	
Other	3	o	1	0	
Not known	1	_34	2	_0	
Total	183	169	134	16	

Table VI-2 gives the percentage distribution of offenders by race (offenders with race not indicated were not included in these calculations). This table shows a distribution of offenders by race that is fairly consistent across ADW, car theft, and robbery. For rape, however, there is a statistically significant higher percentage of offenders of Mexican extraction, accompanied by a smaller percentage of blacks.



Table VI-2

DISTRIBUTION OF PERCENTAGE OF OFFENDERS
BY RACE AND MOST RECENT OFFENSE

		Most Recent	Offense	2.2
Race	ADW	Auto Theft	Robbery	Rape
White	11.0%	12.6%	11.4%	12.5%
Black	82.4	81.5	84.1	68.8
Mexican	2,2	3.7	3,8	18.8
Others	4.4	2.2	0.8	0.0

The distribution of offenders by sex is given in Tables VI-3 and -4. An interesting observation that can be made from these tables is the significantly higher participation of females in ADW than in car theft or robbery. Females account for almost 20% of the ADW offenders and less than 10% of the robbery or car theft offenders.

Table VI-3

DISTRIBUTION OF NUMBER OF OFFENDERS BY SEX

AND MOST RECENT OFFENSE

•		Sex	
Most Recent Offense	<u>Male</u>	<u>Female</u>	Not <u>Indicated</u>
ADW	147	35	1
Auto theft	120	14	35
Robbery	123	10	1
Rape	16	0	0



Table VI-4
DISTRIBUTION OF PERCENTAGE OF OFFENDERS
BY SEX AND MOST RECENT OFFENSE

Most Recent	Sea	ς*
Offense	Male	<u>Female</u>
ADW	80.8%	19.2%
Auto theft	89.6	10.4
Robbery	92.5	7.5
Rape	100.0	0.0

^{*} Offenders with sex not indicated were not included in the percentage calculations.

2. Number of prior offenses. Over 80% of the offenders in each of the four most-recent-offense classifications had been charged with one or more prior offenses. The percentages are given in Table VI-5. The rape and car theft categories had a higher percentage of repeat offenders than did the robbery and ADW categories.

Figure VI-1 shows the percentage of offenders having at least a given number of prior offenses. Four curves are given on Figure VI-1, one for each type of recent offense. For example, the figure shows that 40% of the ADW offenders had six or more prior offenses.

Table VI-6 summarizes the number of prior offenses for repeat offenders. In each group, some had only one prior offense, but some had at least 20 prior offenses. In addition to the minimum and maximum number of prior offenses committed by the offenders in each of the four groups, Table VI-6 gives the average number of prior offenses committed by the persons in each group. The standard deviation is a statistical measure



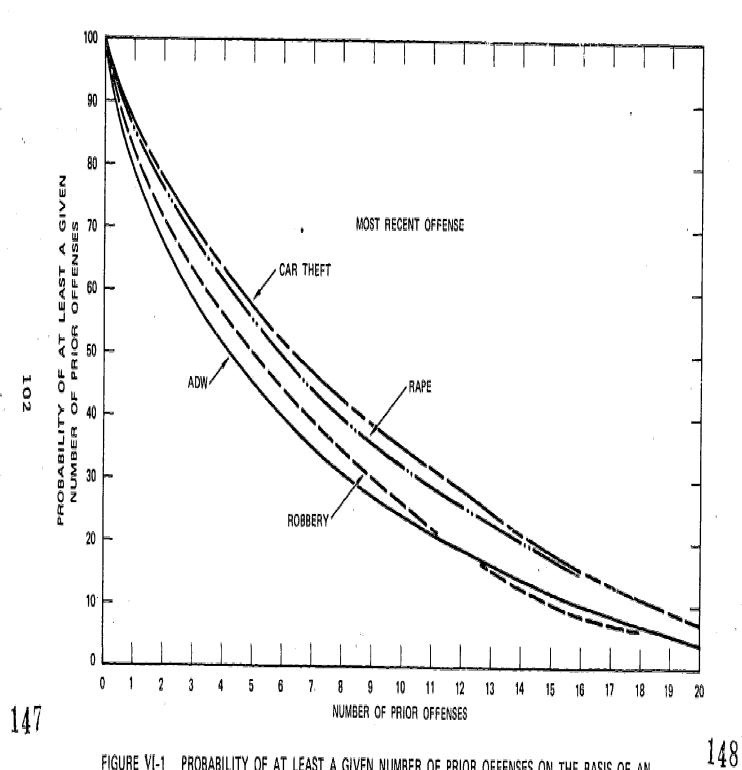


FIGURE VI-1 PROBABILITY OF AT LEAST A GIVEN NUMBER OF PRIOR OFFENSES ON THE BASIS OF AN OFFENDER'S MOST RECENT OFFENSE



Table VI-5

PERCENTAGE DISTRIBUTION OF OFFENDERS
HAVING ONE OR MORE PRIOR OFFENSES

Number of Most Recent Prior Offenses Offense None One or More ADW 19.8% 80.2% Auto theft 14.2 85.8 Robbery 18.7 81.3 Rape 12.5 87.5

Table VI-6

CHARACTERISTICS OF NUMBER OF PRIOR OFFENSES FOR REPEAT

OFFENDERS CLASSIFIED BY MOST RECENT OFFENSE

Most Recent Offense	Average	Minimum Value	Maximum Value	Standard Deviation
ADW	7.3	1	>20	5.9
Car theft	8.8	1	>20	6.1
Robbery	7.7	1	>20	5.7
Rape	9.2	1	>20	6.3

of the variability of the data within each group. If the data are normally distributed, the average plus and minus one standard deviation will enclose about 67% of all the data observations. The average numbers of prior offenses varied between 7 and 9. The ADW and robbery offenders averaged about 1.5 fewer prior offenses than did the car theft and rape offenders. This difference is statistically significant.



3. Types of prior offenses. The hypothesis examined was that repeat offenders in our felony categories would show different patterns of past offenses. Table VI-7 was developed to test this hypothesis. It shows the percentage of offenders, classified by most recent offense, who had at least one prior offense of a specified classification (e.g., of the offenders last charged with ADW, 21.4% had at least one prior felony assault charge).

Some interesting observations drawn from this prior-offense analysis are given below, according to most recent offense:

- ADW. These persons had the highest percentage of past felony assaults. They also showed a high past incidence of burglary, other theft, narcotics and drugs, vehicle law violations, and other crimes.
- Robbery. These persons showed a high past incidence of burglary, auto theft, other theft, narcotics and drugs, vehicle law violations, and other crimes.
- Car theft. These persons had the highest percentage of past car theft, shoplifting, other theft, and stolen propperty. They also showed a high past incidence of burglary, narcotics and drugs, vehicle law violations, and other crimes.
- Rape. These persons had the highest percentage of past burglary, rape, narcotics and drugs, vehicle law violations, and other crimes. They also showed a high past incidence of car theft.
- 4. Age at first offense. Table VI-8 summarizes the age at first offense for repeat offenders. (This table does not imply that the first offense was the same type as the most recent offense.) The average age

Table VI-7
OFFENDER PRIOR OFFENSE PERCENTAGE DISTRIBUTION

		Most Recent		
Prior Offense	ADW	Car Theft	Robbery	Rape
None	19.8%	14.2%	18.7%	12.5%
Strongarm robbery	12.1	12.4	14.9	12.5
Armed robbery	2.8	5.3	10.5	12.5
Felony assault	21.4	14.8	13.4	18.8
Burglary	28.6	47.3	46.3	56.3
Car theft	14.3	40.8	22.4	25.0
Homicide, willful	2.8	1.8	1.5	0.0
Forcible rape	2.2	4.7	2.2	12.5
Attempted rape	0.6	0.0	2.2	6.3
Theft, person	0.6	1.2	3.7	6.3
Theft, purse snatch	1.7	4.1	3.0	6.3
Theft, shoplifting	11.0	21.3	9.7	12.5
Theft, other	28.6	47.9	38.1	31.3
Narcotics and drugs	22.5	29.6	29.9	43.8
Stolen property	7.1	21.3	9.0	12.5
Vehicle law violation	32.4	32.0	23.1	43.8
Other violation	64.3	70.4	58.2	75.0
Other, not indicated	2.8	0.6	11.2	0.0

at the first offense was about 3 years less for the car theft and robbery groups than for the ADW and rape groups. These differences are statistically significant. Table VI-8 also shows the remarkably early age (4 to 11 years) at which some of the offenders began their involvement in crime.



Table VI-8

CHARACTERISTICS OF AGE AT FIRST OFFENSE FOR REPEAT OFFENDERS CLASSIFIED BY MOST RECENT OFFENSE

	Ag	e at First	Offense (year	s)
Most Recent Offense	Average	<u>Minimum</u>	<u>Maximum</u>	Standard Deviation
ADW	19.0	7	43	6.7
Car theft	15.8	5	35	4.9
Robbery	15.2	4	27	4.6
Rape	18.3	11	35	7.1

5. Time from first offense to most-recent offense. This analysis covered the length of time during which the offenders were known to have been associated with crime, that is, the time span between their first and most recent offenses. The data for this analysis are summarized in Table VI-9.

Table VI-9
.
TIME SPAN OF CRIMINAL ACTIVITY FOR REPEAT OFFENDERS

Number of Years from First Offense

to Most Recent Offense					
			Standard		
Average	Minimum	Maximum	Deviation		
10.8	< 1	50	9.7		
7.4	< 1	35	6.6		
8.3	< 1	32	6.6		
12.1	< 1	38	11.1		
	10.8 7.4 8.3	Average Minimum 10.8 < 1 7.4 < 1 8.3 < 1	Average Minimum Maximum 10.8 < 1		



The repeat offenders in the four categories had average criminal records covering approximately 7.4 to 12.1 years. The persons with the most recent offense of car theft had the shortest period of crime (7.4 years). This is significantly less than for the ADW (10.8 years) and rape (12.1 years) offenders.

Figure VI-2 shows the percentage of all offenders in the robbery, ADW, and car theft groups having criminal records covering at least a stated number of years. The rape group is not shown on this figure because of the small number of rape offenders for whom age data were available. For example, Figure VI-2 is read as follows: 20% of the car theft offenders had a criminal record spanning 11 years or more.

offenses per year from first offense to most-recent offense might be considered a measure of the degree of a criminal's participation in crime. Actually, it is only a measure of the number of times that the offender has been apprehended and can be assumed to be a measure of degree of participation only if there is a positive correlation between apprehension and participation. The data summarizing the average number of offenses per year for repeat offenders are given in Table VI-10. This table shows that the persons with the most recent offense of car theft had the highest average offense rate (1.8 per year). This is significantly higher than for the ADW and robbery groups, which had an average offense rate of 1.2 per year. In Table VI-9 it can be seen that the car thieves showed the lowest average number of years in crime.



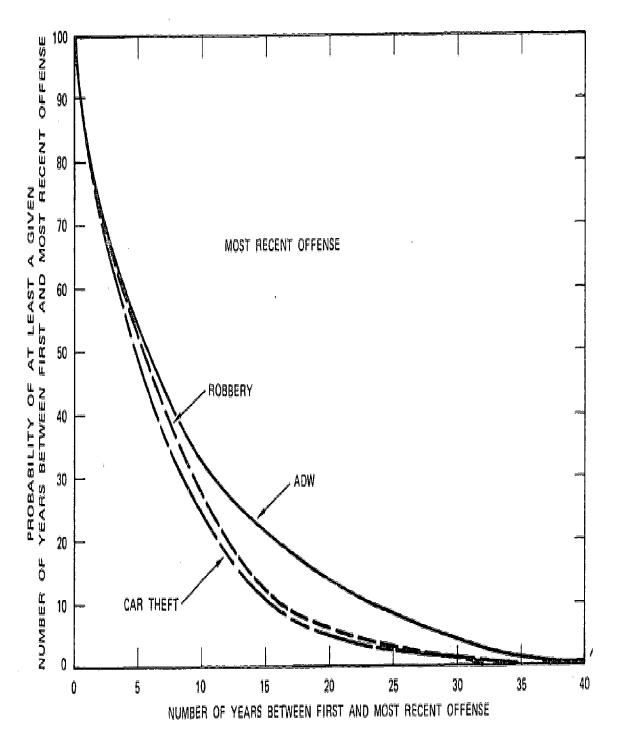


FIGURE VI-2 PROBABILITY OF AT LEAST A GIVEN NUMBER OF YEARS BETWEEN AN OFFENDER'S FIRST AND MOST RECENT OFFENSE BASED ON THE MOST RECENT OFFENSE



Table VI-10

REPEATERS' NUMBER OF OFFENSES PER YEAR

Number of Offenses per Year Most Recent Standard Offense <u>Minimum</u> Deviation Maximum Average ADW 0.0 6.0 1.0 1.1 Car theft 0.3 5.1 1.2 1.8 0.2 5.5 1.1 Robbery 1.3 Rape 1.7 0.3 5.0 1.3

CHAPTER VII. IMPLICATIONS OF UNIFORM DESCRIPTORS FOR INVESTIGATIVE APPLICATIONS

A. Is There a Case for Automated M.O. Investigative Aids?

Many law enforcement agencies have spent hundreds of thousands, and collectively perhaps tens of millions, of dollars in installing a variety of computers to assist in offender identification. The mystique of the computer as manipulating vast amounts of data and spewing out all sorts of information has captured the imagination of hard-pressed law enforcement agencies seeking assistance in tracking and identifying felony crime offenders. However, such systems have yet to demonstrate marked success, particularly in solving modus operandi (M.O.) investigation problems.

A report published in 1972 predicted, on the basis of a survey conducted by the ICMA, that, although "the use of the computer for criminal investigation and dispatch has received little attention to date ... this will change in the future ... Applications for investigations will more than quadruple, rising from 3.7% to 8.8% of the average total police computer use." The report continued that "police ent and resource allocation was clearly regarded as the most important computer use, with crime related files (used for investigation and analysis as well as for reporting) and police patrol and inquiry as second and third, respectively."

The same report went on to state that:

... the surface has only been scratched when it comes to the use of the computer for criminal investigation Several police departments indicated that they had had considerable



[&]quot;Use of Computers by Police: Patterns of Success and Failure,"
International City Management Association (ICMA), Washington, D.C.
(April 1972) pp. 6, 7, 9.

success in automating their field interview reports. Also, there are several experiments underway to establish modus operandi files for use of the computer in tracing criminal patterns and in linking crime to known offenders. (The feelings of the law enforcement community seem split, though, over the utility of modus operandi efforts.)

Although it is still early to make firm predictions, it is quite possible that the computer will have a major influence on the police investigative function. For example, various police officers talked of using the computer to assign cases to investigative officers on the basis of the probability of cases being solved [emphasis supplied] and of the constant interaction between man and machine at all phases of the investigative process.

Certainly the burglary and robbery case follow-up predictive models reported in Chapter II demonstrate the feasibility of implementing the concept of case assignment on the basis of probability of solution, but police investigators must recognize that only a small number of information elements are crucial. This finding may incur anathema from several notable police agencies that have gone to great lengths to attempt to capture vast amounts of personal appearance and M.O. information in anticipation of increasing the likelihood of offender ID and apprehension.

In effect, the results of our research have posed three crucial questions that should receive serious attention by police investigators and planning and funding agencies in their quest for investigative aids:

- What elements of information can police investigators realistically expect to obtain regarding a crime event and the personal characteristics of the offender?
- What are the best procedures for establishing and preserving a logically structured data base that can recall this information in a manner that will materially assist the investigator in solving a crime?



• Is it realistic to expect that the classical concept of M.O. can be developed for automated data processing systems to enable recognition of a distinctive crime commission pattern exhibited by a given offender?

With regard to the third question, it has been noted that the ICMA report stated that the law enforcement community is divided over this issue.

Our findings on criminal activity patterns further reveal no consistency on the part of offenders: They engage in a multitude of crimes.

As the reader can see in Appendix C, we provided for the collection and analysis of a large number of elements of investigative information including M.O. The categories of information in the data collection form are a composite contraction of information elements printed in precoded formats by such police departments as Los Angeles, Miami, Denver, Detroit, and Kansas City.

A CALSPAN document treported the results of an analysis of what might be termed first-generation EDP M.O. systems used by several law enforcement agencies, notably the Detroit and Kansas City, Missouri, Police Departments and the State of Michigan. Although the CALSPAN findings on the hits obtained by the EDP systems used by these departments demonstrated a useful capability, the researchers hedged on recommending (to the State of New York) a headlong rush to implement such a system:



^{*&}quot;Prescriptive Package, Police Crime Analysis Unit Handbook," U.S. Dept. of Justice, Law Enforcement Assistance Administration, National Institute of Law Enforcement and Criminal Justice, Washington, D.C. (November 1973).

[†] Albert Zavala et al., "Use of Computer-Based Modus Operandi Data Systems," Cornell Aeronautical Laboratory Inc. (now CALSPAN Corp.) Buffalo, New York, October 1970.

Increasing system utilization on 61 documented hits in one department is not sufficient, per se, to recommend the establishment of a full scale M.O. system. However, the reliability of M.O. as an identification tool has been shown ... [and] ... that M.O. data can be used effectively in combination with personal appearance information.

B. The OPD Crime File System

The OPD Crime File System is a known-offender-based system with supporting vehicle and fingerprint subsystems. This system became operational in 1973, following the inputting of selected categories of known felons and their physical appearance characteristics. It provides four main categories of information.

- Physical characteristics of offenders (derived from certain categories of arrest records).
- · The types of crimes that the known offenders have committed.
- Mugshots and fingerprint displays of the subjects in the system.
- Listing the descriptions of vehicles obtained from citations and selected FI reports.

Figure VII-1 illustrates the Subject File Query form, which contains the address codes for data elements entered into the computer subject file input (SFI). Figure VII-2 shows the address codes for the Vehicle File Input (VFI) form. The forms illustrated are used by the Crime Analysis Section (CAS) computer operators to interrogate the Crime File memory bank, using the descriptive information on offenders and vehicles contained in the felony reports that are tagged for enrichment. The data elements contained in the SFI and VFI forms were subjectively selected for the Crime File System by OPD R&D staff, with assistance from the Criminal Investigation Division (CID).

During the course of our analysis of investigative sources of information leading to a felony suspect's ID, we noted on Card 7 of the data



	MMER	(enter cii or opo number)				E QUERY FORM				DATE
	CII CED = AME		LĀSĪ,	TF =5368 First		KKLAIO POLICE DEPARTNENT	i			
	TWO OF	SEX ['CRCLE ONE'] A] Mole A2 Female	INCOME	RACE [CIRCLE ONE OR TWO] B1 White B2 Block B3 Brown B4 Yellow B3 Undetermined		AGE [ENTER MINIMUM & MAXIMUM A C] [** YEARS OF AGE)	NGE LIMITS]	RESIDENCE CODE LENTER UP TO 3 ONC-OIGHT DISTRIP	T NOS.	
H	EQ # EI, EZ EQ Armad ro ET Strongam EQ # E6 or EQ Auto thef EIQ Grand 1 EIL Check o EIZ Felony EIL Macade EIL Rope EIL Child m EIQ Other-w EIL Other-w EIL Other-w EIL Other-w	E OR TWO] of burglary of burglary of burglary on of stolen property 1,53, or E4 obblary on and purse enotich of E7 in theft ons crudit card ossault or E13 d exposure notesting ex	EL WEIGHT MAXIMUM V G1 L/ HAIR CO	[ENTER MINIMUM AND VEIGHT LIMITS] POUNDS] LOR E OR TWO] OF GREY	M3 Excess M4 Mole 1 LIPS A [CIRCLE N1 Hoirip N2 Unusus N3 Other COMPL [CIRCLE P] Light, P2 Oark,	ONE] Over or missing sively protruding with conting(s) ONE] ONE Democratic deformity EXION ONE OR TWO] Indir black and or sphotchy	[CIRCLE T] Arms T2 Hondi T3 Lept T4 Other VISIBLI MARKS CGIRCLE U2 Hondi U3 Foca, U4 Other SPEEC [CIRCLE VI Forial	is or lingers or feet or combinations of the above E SCARS, MOLES, BIRTH- S, OR NEEDLE TRACKS A ONE] is is or fingers in head, or neck or location or combinations of the above OH A ONE OR TWO] ago occent applied registral occent	[CII W] W2 W3 W4 NIC	CULIARITIES & COLE ONE] Limp Effeminate male or mancular female When chithing of opposite sen (impersonator) Twitch of eyels), foce, or other KNAME TER UP TO 10 CHARACTERS, IF C, ENTER FIRST NAME]
	E22 Anon E23 Multiple E24 Operate such as E25 Associat E26 Associat E27 Associat E28 Associat E28 Associat E29 Associat E29 Associat	I felonies In learns (major crimes only, to robbery and burglary) ted with sheplifting ring ted with check/tradit cond ring ted with group advocating or proc- riolence ted with gestortion/loan shorts ring ted with consumer troud/burco ring	Ji Bald Ji Portly Ji Close c Ji Medium Ji Long (I Ji Alro on Ji Alro on	length below ears) nericon-netural style nericon-processed style wavy, or kinky	[CIRCLE OI Arms O2 Hands O3 Foce O4 Other FACIAL	or fingers and neck location or combinations of the above HAIR <u>A</u>	Fin	MUG SHOT ADI		
		ted with organized prostitution	EVE CA	I AB A	R <u>i</u> Yes		F	INGERPRINT CODES		

CENTER FINGER NUMBER AND THREE-DIGIT CODE, UP TO THREE TIMES; USE "?" FOR UNKNOWN CHARACTERS)

CRIME SYSTEM



EN Associated with organized prostitution
EN Associated with organized narcotics

E32 Associated with organized gombling

🔁 Associated with outo stripping ring

EYE COLOR A

[CIRCLE ONE]

K<u>I</u> Brown or block

Kg Blue, gray, green, or hozal

LL Either eye blind, missing, or ortificial LZ Wedra glasses (prescription)

EYE DEFECTS A

[CIRCLE ONE]

LŽ LI OF LŽ

TEETH A

(CIRCLE ONE OR TWO)

Si kreqular or protruding

SZ Metal fillings visible

S) Visible decay or stains

54 Folse, chipped, or missing teeth

VEHICLÉ FILE INPUT (VFI) FORM [QUERY ONLY]

DAT	Ē		

FIELD CONTACT
CIRCLE ONE
AL Yes AL No
DATE OF CITATION
ENTER CITATION PERIOD AS "DAY, MONTH,
& YEAR", I OR 2 SIX-DIGIT DATES
Bī
DDMMYY'ODMMYY
YEAR OF VEHICLE
ENTER TWO-DIGIT YEAR NUMBER
%19
MAKE/MODEL
CIRCLE ONE MAKE AND A MODEL IF APPLIC-
ABLE DI Alpha Romeo
DŽ Alpinė
D3 American Motors #Q No specific model
#1 Ambessodor
#2 AMX #3 Gremlin
/4 Hornet
5 Javelin # 6 Rambler American
(See D38-/2/046-/I)
D4 Aston: Mortin D5 Audi
DG Austin
DZ Austin Healey DB Bentley
DO BMW
D <u>IO</u> BMW D <u>II</u> Borgward
D12 Buck ✓Q No specific model
<u>I</u> Riviera
#2 Skylark D13 Cadillac
/ Q No specific model
기 El Dorado (FLE) D년 Capri (import)
D <u>15</u> Chevrolet
/O No specific model /_ Camero
- ₹ <u>2</u> Caprice
/3 Chevelle /4 Chevy ()
/5 Corvair /6 Corvatte
≠ Z El Camino
/g Impole /g Molibu (ELL)
≠ <u>iQ</u> Monte Carlo
/ <u> Nova (CH2)</u> / <u> Vega</u>
D <u>16</u> Chrysler
/Q No specific model / Imperial
DIS Deseto
DI9 Dodge
#Q No specific model #1 Challenger
₹2 Charger
/3 Corpnet /4 Dart
Dune Buggy (See D51-#()

•		
1	MA	KE/MODEL -CONTINUED-
П	AUT	OMOBILES - CONTINUED -
	020	Edsel
Т	D <u>21</u>	English Ford (British)
	052	Ferrari
	23	Fial
	224	Fiat-Abarth
1,	25	Ford
	-	
1	1	
	1	
1		
	,	
1	i	
1	1	
ı	1	Pinto
1	/	
1	-	
١.	- /	
	26	Hillman
	21	Honda
	28	international (Härvestei) Jäguar
	<u>30</u> 5 <u>9</u>	nging.
	31	Kormann Ghia
	32	Lincoln
ľ	70	
l	1	
l	13	
	33	Lotus
	<u>34</u>	Mazda
	35	Mercades-Banz
٥	36	Mercury
l	Ø	
ļ	1	
١,	≠2 37	MG
	38	Nash
٦	<u> </u>	
	1	
	12	
		(See D3-#6/D46-#1)
D,	39	Oldsmobile
	Q	No specific model
_	1	Cutioss (F-85)
	40	Opei
	11	Packard Packard
04		Peugeot Plymouth
-	<u>.</u> •0	
	1	Barracuda
	12	Belvedere
	13	Duşter
	14	Fury
	/ 5	GTX
	1 <u>6</u>	Road Runner
	#7	Satellite
D4	18	Valiant Pontiac
~3	#0	No specific model
	1	Bonneville
	12	Firebird
	/3	Grand Prix
	14	GTO
	13	LeMans
	≠ 6	Tempest
D4		Porsche
04		Rambler
	₽Q <u> </u>	No specific model American
	* ±	terrolet (Appl)

1444	/F /MODEL
	KE/MODEL -CONTINUED-
	OMOBILES - CONTINUED-
0 <u>48</u> 0 <u>49</u>	Rolls Rayce Saab
D5 <u>0</u>	Shelby American
	(See -D25-#7)
D <u>51</u>	Special Vehicle
19	
•	
D <u>52</u>	Studebaker
D <u>53</u>	Subaru
055	Sunberm Suzuki
D56	Toyota
057	Triumph
058	Volkswagan
059	Volvo
D€Ō	Willys=Overland
	PRCYCLES
061	BSA Hada - Bawdaaa
D ē 3	Harley-Davidson Honda
0 <u>64</u>	Kawasaki
065	Sužuki .
066	Triumph
067	Yamaha
BOD	Y TYPE .
CIRCL	
EI	2-door
Ē2	4-door
E3	Station Wagon
E₫	Convertible
E <u>5</u>	Pickup
E6	Sportscar
E <u>7</u>	Van Panel
E <u>8</u> E9	Pus .
E <u>IO</u>	Hearse
COL	
1/5.	ONE (SINGLE COLOR) OR TWO (2-TONE)
F <u>i</u>	<u>Ba</u> F <u>i</u> Block
FŽ	F2 Brawn, Beige, Bronze, or Tan
F3	F3 Red, Pink, or margon
F <u>4</u>	F4 Orange
F5	F5 Yellow or Gald
F <u>ē</u>	F6 Green or Turquoise
F <u>7</u> E 0	F7 Blue
F <u>B</u> F <u>9</u>	F§ Purple or Lavender F§ Silver or Grey
FIO	FIQ White or Cream
*N	OTE:
	T/S = TOP OR SINGLE COLOR
	B=BOTTOM COLOR
STAT	E
IRCLE	
3Ĭ SWOCE	California
32	Other
ICE	NSE NUMBER
	UP TO 6 CHARACTERS LEFT JUSTIFIED
NTH A	LL BLANKS (SPACES) OUTTED AND
NSERT	LL BLANKS (SPACES) OMITTED AND "7" FOR UNKOWN CHARACTERS
ŧĮ.	

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CRIME SYSTEM

FIGURE VII-2 OPD VEHICLE FILE QUERY FORM

(See D3-#6/D38-#2)



collection form (shown in Appendix C) whether Items 3 and 4, Crime File run-person, and Crime File run-vehicle, respectively, had actually been run on the cases we analyzed, and whether the results of the computer file search had been useful. Table VII-1 summarizes the numerical results observed for the sample of cases analyzed.

The Crime File statistical analysis results are not impressive with regard to the number of runs that had yielded useful leads to offenders on the basis of personal appearance descriptions and descriptions of vehicles that might have been involved in a crime. The Vehicle File appears to have produced exceptionally poor results. By comparison, however, vehicle registration check via the Police Information Network (PIN) and the California Law Enforcement Telecommunications System (CLETS) when the registration number was available, in the case of robbery (as shown

Table VII-1

RESULTS OF CRIME FILE SYSTEM UTILIZATION

	. Crime File Run						
	Pe	Person		Vehicle			
				Vehicle			
	Cases	Runs	Cases	Vehicle	Linked to	Runs	
Felony Category	Run	Useful	Run	Stolen	Suspect	<u>Useful</u>	
Robbery						ala.	
(armed & strongarm)	139	3	11	1*	1*	1*	
A = = = 1 +	35	4	9			0	
Assault	33	•	•				
Car theft	6	0	0			0	
D==0	25	1	8	==	~-	1	
Rape	23	-	Ü	11221111		_	

^{*}Same case.



in Table II-17, the case decision model), were much more useful investigative aids.

We found no indication that the Crime File fingerprint subsystem had ever been used for analysis when latent fingerprints had been recovered for the few cases so noted. This subsystem may be more useful for burglary cases, but we did not pursue such an analysis.

In Table II-17, the robbery decision model, where the data category elements have been assigned numerical weights in accordance with the contribution they were found to have made to case, clearance, physical appearance descriptors do not appear as significant. Physical appearance descriptors certainly must make some contribution to case solution, since the Crime File run-person variable carries some statistical weight for robbery cases. The fact that a few useful leads are shown in Table VII-1 reveals that the query of the data bank must have produced a range of mug shots from which a victim or witness identified the suspect.

We cannot give a conclusive explanation for the OPD Crime File System's failure to show a significant contribution to the cases in our study. We did observe one failure obviously attributable to the incompatibility of the SFI descriptors in the computer memory file and the operator's interpretation of the subject descriptors in the incident report. It is apparent that the success of the operation is greatly affected by the way the operator must interrogate the memory bank to accommodate variations in suspect personal appearance characteristics as they appear in the incident report.

In addition to observing the general CAS Crime File and other EDP investigation operations, we observed a test case conducted by the lieutenant in command of the Robbery Section.

The test case produced by the lieutenant was a robbery that had been reported a week or so earlier but had not been processed through the Crime



rile System. At the time the test case was run, the robbery suspect had been arrested for another crime, and the officer wanted to follow the procedures by which the case was processed. These procedures were carried out by the most experienced Crime File operators available.

The height and weight characteristics of the suspect as given in the reported robbery incident were keyed into the computer, with slight adjustments of inches and pounds to allow for variation in the subjective judgments of the victim and the officer. (See Figure VII-1 for the descriptors referred to in this example.) The computer is so programmed that precise measurements and descriptors must be keyed in for the initial pass. The search process automatically suppresses possible suspects whose descriptors are at variance with those keyed in.

A major problem was encountered in the keying of the suspect's hair color. The incident report stated that the suspect had long "reddishblond" hair. Note that the codes for hair in Figure VII-1, H1 through H5, allow only for blond or red hair--not reddish-blond. Consequently, the operator keyed in long blond and long red hair. Obviously, judgment as to what constitutes long hair is somewhat subjective.

Another problem concerned the type of crime committed, so that the data bank could be queried as to the prior-crime M.O. of the suspect. The robbery incident was coded as armed robbery, because, when the suspect had been surprised in the act of committing a burglary, he drew a gun and fled. Since the report was classified as a 211 P.C. (robbery penal code) and not with an additional 459 P.C. (burglary), the operator made the error in judgment of keying only Code E6 (armed robbery). The operator also overlooked the fact that the suspect had been reported to have a tattoo (Codes Q1 through Q4). Together the descriptors produced much too large a "hit" range for the first pass.



5.5

9.

Before proceeding further, the name of the suspect was given to the operator to determine whether this name was in the system. The operator keyed in the OPD arrestee file number, the State Department of Justice CII number, and the suspect's name. The correct identity and mug shot were produced. The subject was shown to have a prior record of burglary, strong-arm robbery, drug, and vehicle violations.

After it had been determined that the suspect's name was indeed in the system, a more carefully considered set of descriptors was keyed, including multiple felonies, Code E32, and facial Code R1 (the report had noted a "droopy" mustache). The teleprinter produced the name of five subjects matching these descriptors. The suspect's name was among the five printed out. A request was made for the five mugs to be shown. Incredibly, the CRT console failed to retrieve the named suspect's mug. An error was then discovered in the address coding of the suspect mug to his SFI record.

Coincidentally with our inquiries regarding the Crime File System's results, the OPD discovered basic flaws in the system that could not be immediately explained. A total core dump was made, and each entry was 'carefully checked. One problem was traced to a crack in the optical lens system; another was attributed to the software program. The OPD is of the opinion that the supplier failed to deliver a fully operational system. The software system underwent redesign by an OPD consultant software specialist. We understand that the OPD believes it has now corrected the problems.

C. Comparison of Alternative M.O. Systems

In retrospect: What have we learned about a computerized suspectoriented file system? Although the many agencies that use such systems can cite anectodal examples of successes achieved—and probably achieved



when other approaches would have required an enormous effort to search files—the simple fact seems to be that the successes are not spectacular in a statistical sense. The OPD Crime File System, as contrasted with others, appears to have been designed with the approach that the criminal population with which the OPD has to deal is largely a recidivist population. Chapter VI has shown that over 80% of the offenders whose criminal histories were analyzed had had prior offenses. The repeat offenders averaged more than seven prior offenses during their span of contact with the OPD (or other agencies).

A major police department has had under development since 1969 an extremely complex pattern recognition and information correlation system. Originally, a number of subsystem capabilities were envisioned. But with the passing of time, the system design was scaled down to provide three basic information subfiles, on persons, events, and vehicles. The input data are to be derived from incident and field interrogation (FI) reports as they occur. As of the summer of 1975, the system was not known to be operational, despite extensive field testing since 1971. We mention this system and one other mainly to contrast their data input structure with the OPD's less ambitious Crime File System. No event information is stored in the OPD system.

Two police agencies' M.O.-type systems that we have looked at have been designed to be built up and kept current on the basis of events and associated suspects. Consequently, both departments will be inputting, to the computer storage system, suspect information developed by as many officers and investigators as generate the reports. Both systems, as contrasted to the OPD, will input data for unsolved cases, with the desired objective of developing patterns of events as well as the M.O. of suspects. Much as our technological imagination would like to see such systems succeed, we cannot overlook their relative inability to live up to the expectations in the past, for very fundamental reasons.



Figures VII-3 and VII-4 show descriptor and M.O. checklist extracts from two police agencies' incident reports. The checklist forms were designed for use by patrol officers and possibly by follow-up investigators. The extremely detailed personal appearance and crime attribute data were intended for input into the central data processing center. The data sought are not unlike the type of information we were seeking in the OPD reports. Although we concede that the OPD patrol is not provided with such extensive and detailed check lists to query witnesses and victims, we seriously doubt that victims who have been subjected to the trauma of an armed robbery or assault can respond well to extensive questioning as would be required by these represented forms. incident report form asks for general characteristics of the crime and the perpetrator. We found, consequently, extreme variation in the level of details recorded. But there was no way for the analyst-coders to ascertain whether the limiting factor was the victim (other than a statement that, in the officer's opinion, the individual reporting was under the influence of drugs or alcohol or was injured). Whether the reporting officer had even bothered to try to secure certain information could not be ascertained from the reports analyzed.

We can state with some confidence that, because of the sample size we drew, and the fact that OPD patrol officers are probably sufficiently trained to ask for and record vital information, it may be unrealistic to expect officers to obtain such fine details at a crime scene as are indicated in Figures VII-3 and 4. Also, this level of detail has yet to be demonstrated as generally useful. Although the information may be available, the cost in time and resources to collect and process the data will be high. Furthermore, we strongly suspect that patrol officers would find the filling out of such forms burdensome and too time-consuming, unless they had been convinced of benefits.

	1 2 3 303 Tattoo (cont.	307 Teath	311 Eyes	314 Facial Hair	317 Face
TIES	3 0 Amputee 3 3 3 NAMES 4 4 4 WARDS 5 5 5 INITIALS 6 6 6 6 PALPHUCO 5 5 5 EAR 304 Facial Scars	1 1 1 MISSING 2 2 2 GOLU 3 3 3 BHOKEN 4 4 4 FALSE 5 5 5 STAIN DECAY 6 6 6 PROTHUDING 7 7 7 IHREGULAR 308 Body Scars	1 1 1 MISSING 2 2 2 CHOSSED 3 3 3 SUNCLANSES 4 4 4 CLASSES IPLAINI 5 5 5 BULGING 6 6 6 SQUINTIBLING 7 7 7 SLANTED	1 1 MUST = CHINESE 7 2 2 GOATEE 3 3 3 BEARD = FULL 4 4 MUST = HEAVY 5 5 5 MUST = THIN 6 6 6 MUST = MEDIUM 7 7 7 BROWS = HEAVY 8 8 8 UNSHAVEN	1 1 NEGHO W CAUC. FEATURES 2 2 2 HI CHER BONE 3 3 3 LONG 4 4 4 BHOAD 5 5 5 THIN 6 6 6 HOUND
NAL ODDITIES	301 Deformed 2 2 2 CHIN 3 3 3 FOMEMEAD 4 4 4 LIP 2 2 2 ARM 5 5 5 MOSE 3 3 3 HAND 7 7 7 EYEHROW 5 5 5 FINGERS	1 1 1 ARM 2 2 2 HAND 3 3 3 WRIST 4 4 4 NECK 5 5 8 BUNN 6 6 6 CHEST	312 Hair Type 1 1 1 DVED 2 2 PHOCESSED 3 3 3 WIGGROUPE 4 4 CHEW CUT	315 Ears 1 1 1 CAULIFLOWER 2 2 PIERCED 3 3 3 PHOTHUDING 4 4 4 CODE TO HEAD 5 5 5 LARGE	318 Complexion 1 1 1 DARA 2 2 2 SALEDW 3 3 3 RUDDY 4 4 4 LICHTI-FAIR 5 5 5 MEDIUM
PERSONAL	6 6 SOWLEGGED 305 Facial Oddity	309 Speech	5 5 8 BALO 6 6 6 AFRO 7 7 7 LONG 8 8 THIN/RECEDED	5 6 6 SMALL	319 Other
	1 1 1 POCKMARKS 1 1 1 1 POCKMARKS 2 2 2 MOLES 2 2 2 MOLES 3 3 3 FREGRES 4 4 4 PIMPLES 4 4 4 CHESTINECK 5 5 5 LIPS - THICK 6 6 6 6 LIPS - THICK 7 7 7 CHIN-PROTHUDE 1 1 1 PICTURES 8 8 8 CHIN-RECEDES 2 2 2 DESIGNS 9 9 9 HOLLOW CHEEK	1 1 1 ACCENT (U.S.) 2 2 2 ACCENT (OTHER) 3 3 3 USPS 4 4 4 STUTTERS 5 5 6 HARE LIP 6 6 6 MUMBLES	9 9 9 STRAIGHT 313 1 1 1 WAVY 2 2 2 BUSHY 3 3 3 CURLY	1 1 CHOOKED 2 2 HOOKED 3 3 3 UPTURNED 4 4 LONG 5 5 5 BROAD 6 6 6 FLAT 7 7 7 SMALL 8 8 8 THIN	1 1 1
WEAPON	401 0 0 0 GUN 5 5 5 6 INCH OR MORE 1 1 1 REVOLVER 6 6 6 8 BLUE STEEL 2 2 2 AUTOMATIC 7 7 7 NICKEL PLATED 3 3 3 2 INCH 8 8 8 NUNUSAL GRIPS 4 4 4 4 INCH 9 9 9 RUSTY/DEFECTIVE	Caliber 40 20 RARE 21 RIFLE 22 SHOTC 23 SAME 3 24 MACH	26 OTHER GUN SUN 27 SIMULATED GUN		48 RAZDR 46 BRASS KNUCKLES

	26 Misc. Sex Act 30 RAPE OR ATTEMPT 37 OTHER:	22 Pretend (cont.) 35 REPAIR/DELIV'MAN 39 CUSTOMER 40 OTHER:	24 Solicit (cont.) 32 INFO 21 A CHAPETTE 39 OTHER	22 Vehicle Involved 77 HID IN REAR SEAT 80 V. FORCED INTO VEH, 81 VICT. PKG., GARAGING 75 FORCED VICTIMS	25 Force (cont.) 48 CHOKED 50 KICKED 49 HIT (OTHER) 52 ROLLED 64 THEOTER TO KILL	21 Statements 81 APPLICATION 84 EXTRAORDINARY 25 Carried Gun In
RKS	24 Suspect Wore 10 MASK/FAGE COVER 11 BAG, CLDTH, E' WEYEROLES 12 HALLOWSEN 13 HANDKERCHIES SCARF ON FACE 14 SKI MASK 15 STOCKING	23 Susp's, Actions 36 USED NOTE 25 HID IN BLDG, 24 DEMANDED MONEY FROM SAFE 27 JUMPED COUNTER 15 DEMANDED #WELRY	24 Telephone 84 CONTACTED BY 81 PULLICUT/DISCONNECT WHES 90 OTHER:	73 VEHICLE TO CURB 73 DISABLED VS. VEH. 72 COVER, BEND, ALTER, LICENSE PLATE 79 STOP VS. VEH. BY FLAGGING DOWN ETC, 76 FORCED WAY INTO VEH. 83 OTHER:	70 OTHER: 25 Vict, Forced To 11 DISROBE 12 ENTER VEH, TRUNK 14 ENTER REFRIG.	96 SHOULDER HOLSTER 91 BAG/BRIEFCASE 92 WAISTBAND 93 NEWSPAPER 94 POCKET 97 OTHER
TRADEMARKS	16 OTHER MASK 03 CLOTHES OF OPP. SE 06 MAKE UP IMALES ON 05 GLOVES 04 UNUSUAL CLOTHES 02 CAPMAT	34 USED LODKOUT 35 USED DRIVER 43 OTHER 24 Solicited/Offered 23 AID FOR VEHICLE	25 Evidence 22 LEFT NOTE 21 PRINTS-AVOID/ REMOVE 28 FINGERPRINTS .30 OTHER	25 Force 61 KIDNAPED 67 TORTUNED 54 HANDCUFFED 53 GAGGED	17 REAM OF BLDG. 15 OPEN SAFE ROOM 18 ENTER REST ROOM 13 I.IE ON FLOOR 15 OPEN REGISTER 19 OTHER.	21 Type 26 HIJACK 27 HOMOSEXUAL 36 PAYROLL HOLDUP 40 SMATCHED 41 STRONGARM ,44 OTHER.
	22 Pretended To 31 POLICE 33 AIOING VICTIM 38 BLIND, CRIPPLED, INFIRM, ETC. 37 SEEKING SOMEONE 42 RENTING	9 36 AIDE 27 LIOUDH-BRAND 22 CIGARETTES-BRAND 28 NARCOTICS 31 IMMORAL ACT 38 USE PHONE/TOILET	22 Shots Fired 93 AT VICTIM 97 WARNING 92 AT POLICE 91 ACCIDENTAL 98 OTHER	45 BOUND 45 THREATEN V'S FAMILY 43 BLINDFOLDED 53 SEARCHED 56 HIT W/WEAPON 42 BIT 51 CUT/STABBED	22 Initial Contact 01 AMBUSHEO 05 ATTACKED FROM REAR 12 RANG/KNOCKED 24 OTHER	22 Victim Was 56 OPENING/CLOSING 62 AGED, BLIND, CHIPPLED, ETC, 62 OTHER.

FIGURE VII-3 ROBBERY REPORT CHECK-OFF LIST

PG OF	NOSE			
MALOR OF THE PROPERTY AND ADDRESS.	i∐nknown 1 2 3	SPEECH Unknown 1 2 3 Accent/for, Lang. 1 2 3	Left Tools 1 2 3 Used Vic. Tools 1 2 3	C. R. N. SUSPECT'S INJURIES
SUSPECT INFORMATION	Small 1 2 3	Accent/For. Lang. 1 2 3	Prepared Exit 1 2 3	None Visible 1 2
EACE SHAPE	Medium 123 Large 123	Impediment 1 2 3 Mumbias 1 2 3	Rensacked 1 2 3	Visibly Upset 1 2 Miner Injury 1 2
Unknows 1 2 3 Heart 1 2 3	Thin t 2 3	Pitch. High 1 2 3	Was Tidy 1 2 3 Selective in Items 1 2 3	Serious Injury 1 2
Oval 1 2 3	Broad 1 2 3		Took Items in	Death 1 2
Round 1 2 3	Long 1 2 3 Flat 1 2 3	Speed, Fast 1 2 3 Slow 1 2 3	Vic. Container 1 2 3	Unknown 1 2 ;
Squara 1 2 3	Crooked 1 2 3	Valume, Soft 1 2 3	Used Trailer 1 2 3 Bound Victim(s) 1 2 3	VICTIM'S INJURIES None Visible 1 2
FACIAL GODITY	Crooked 1 2 3 Hooked 1 2 3 Upturned 1 2 3	Loud 1 2 3	Covered Victim 1 2 3	Visibly Upset 1 2
Unknown 1 2 3		SUSP. EMOTIONAL STATE Unknown 1 2 3	Disrohed Fully 1 2 3	Minor Injury 1 2
Birthmarks 1 2 3 Fracklas 1 2 3	TEETH Unknown 1 2 3	Annry 1 2 3	Disrobed Partially 1 2 3 Fondled Victim 1 2 3	Serious Injúry 1 ? Death 1 2
Moles/Warts 1 2 3	MAGE ' 2 1	l Palm 1 9 3 1	Masturbated 1 2 3	Unknown 1 2
Pimples 123	Broken ULUL	Determined 1 2 3	Penetrated Victim 1 2 3	
Pockmarks 1 2 3	Gepped ULUL Irregular LULUL	Excited 1 2 3	Perl. Act of Sodomy 1 2 3	VICTIM'S ACTIONS None/Unknown 1 2
FACIAL SCARS	Protruding VLULUL	Irrational 1 2 3 Nervous 1 2 3 Stupor 1 2 3 Other #1 Other #2 Other #2	Climexed 1 2 3 Used Prophylectic 1 2 3	Attowed Entry 1 2 Apprehending Susp. 1 2
Unknown 1 2 3	Protruding ULULUL Missing ULULUL Stain/Decay ULULUL	Stupor 1 2 3	Used False ID. 1 2 3	Apprehending Susp. 1 2 Arguing W/Susp. 1 2
Forehead RLRLRL	Stain/Decay ULULUL Bridge ULULUL	Other #2	Used Sympathy 1 2 3 Mede Promise(s) 1 2 3	Bystander 1 2
Eyebrow RIRLRI Nose RIRLRI	Partiat ULULUL) Usher #3	Limped 1 2 3	Committing Crime 1 2
Charle Di Di Di		SUSP. PRETENDED TO BE	Bod Madical Card 1 1 1	Customer/Client 1 2 : Engaged in Fight 1 2 :
Ear RIRIRL Lip ULULUL Chio RIRIRL	Capped, Gold ULULUL Silver ULULUL	Pandus Cuspay 1 1 1	FORCE USED BY SUSP. Unknown 1 2 3	Escape 1 2
LIP ULULUL Chia RIBIBI	Designed ULULUL	Crippled 1 2 3	Unknown 1 2 3	Interfering 1 2
	OTHER FACIAL	10	Cuttina 123	Intox./Drugs 1 2 Oper. Motor Yeh. 1 2
FACIAL HAIR Unknown 1 2 3	Unknown 1 2 3 Lips, Thick 1 2 3 Thin 1 2 3	Deliv. Person 1 2 3 Disabled Motor. 1 2 3		Part, in Disturb. 1 2
Renne Thin 1 2 3	Lips, Inick 1 2 3	Disabled Motor. 1 2 3 Employee 1 2 3	Shooting 1 2 3 Shoving 1 2 3	Pedestrian 1 2
Madium 1 2 3	l Chin. Protrude 123	Emptayer 1 2 3	Striking 1 2 3 Striking-Wespon 1 2 3	Sleeping 1 2
Bushy 1 2 3	Recede 1 2 3 Double 1 2 3	friend	Threalening 1 2 3	Vehicle Accident 1 2 Other 1 2
Madium 1 2 1	Dimale 1 2 3	III 1 Z 3 I Injured 1 2 3	MEAN DIGED BY CHES	VICTIM FORCED TO
Bushy 1 2 3	Chask Dimple RLRLRL	Need of Assist. 1 2 3	MEAR'S USED BY SUSP. Unknown 1 2 3	Commit Crime 1 2 Consume Lig./Pill 1 2
Beard, Full 1 2 3 Ear Hair 1 2 3 Gostee 1 2 3 Nose Hair 1 2 3	BODY SCARS Unknown 123 Neck RIRIR Shoulder RIRIR Upper Arm RIRIR Etbow RIRIR	Need Phone 1 2 3 Police 1 2 3	Pistot U	Consume Lig./Pill 1 2 Disrobe 1 2
Goates 1 2 3	Neck RIRLRI	Public Utit. 1 2 3	Revolver """ U	Disrobe Partially 1 2
Nose Hair 1 2 3	Shoulder RLRLRL	Relative 1 2 3	Ritle Sawed 1 2 3 U	Drive Motor Veh. 1 2
Sideburns, 1 2 3 Mid. Ear 1 2 3	Upper Arm RLRLRL	Renting 1 2 3	Shotgun 1 2 3 U	Enter Building 1 2 : Enter Restroom 1 2 :
HATOW I Z J	llower Arm RIRLRI	Sales Person 1 2 3	Shotgun, Sawed 1 2 3 U	Enter Room 1 2
Unshaven 1 2 3	Wrist RLRLRL	Seeking Someone 1 2 3	Other Gun 1 2 3 U	Enter Vehicle 1 2
HAIR TYPE		Soliciting 1 4 3	Brand #1	Enter Veh. Trunk 1 2 : Leave W/Suspect 1 2 :
Unknown 1 2 3	*Fingers RLRLRL Chest RLRLRL Abdomen RLRLRL	Other #2	Brand #7	Lie on floor 1 2 :
Afro 1 - 2 2 - 3 Afro, Mini 1 2 3	Abdomen RLRLRL	Police 1 2 3 Public Util. 1 2 3 Relative 1 2 3 Renting 1 2 3 Renting 1 2 3 Sales Person 1 2 3 Sales Person 1 2 3 Soliciting 1 2 3 Soliciting 1 2 3 Other #1 Other #2 Other #2	Brand #1 Brand #2 Brand #3 Caliber/Ga. U	Open Cash Register 1 2
Baid 1 2 3	Back RLRLRL Side RLRLRL	Jouann Tune	nunantinuka i ta ö	Open Safe 1 2 Perform Sex Act 1 2
Bald 1 2 3 Braided 1 2 3	Butlocks RLRLRL Upperleg RLRLRL Knee RLRLRL	Unknown 1 2 3 Contacts 1 2 3	Blue Steel 1 2 3 U Nickel Plated 1 2 3 U	Rear of Bidg. 1 2 3
Crew Cut 1 2 3 Curly 1 2 3	Upper Leg RLRLRL	Contacts 1 2 3 Glasses 1 2 3	Rusty 1 2 3 U	Use Telephone 1 2
Dyed/Bleached . 1 2 3	Knee RLRIRL Lowerleg RIRLRI		**************************************	Other 1 2 :
Fronted 1 2 3	Ankle RLRLRL	Blande 1 2 1	Switchblade 1 2 3 U Razor 1 2 3 U	Unknown 1 2
Natural 1 2 3 Processed 1 2 3	FOOT REALES	Sun 1 2 3	Chapping Tool 1 2 3 U	Cutting 1 2 3
Straight 1 2 3	DEFORMED Unknown 1 2 3	Wire frame 1 2 3	ice Pick 1 2 3 U Surgicel Inst. 1 2 3 U	Pulling 1 2 3 Sheeting 1 2 3
Streeked 1 2 3 Thin/Receded 1 2 3	Uppar Arm RLRLRL	Mask 1 2 3	Bottle 1 2 3 U	Shoving 1 2 3
Thin/Receded 1 2 3 Wavy 1 2 3 Wig/Toupe 1 2 3	lower Arm RIRIRI	Hearing Aid 1 2 3	Club 1 2 3 U	Striking 1 2 3
Wig/Toupe 1 2 3	Hand RLRLRL *Fingers RLRLRL **RETTER RETTER	Wire Frame 2 3 Gloves 1 2 3 'Mask 1 2 3 Hearing Aid 1 2 3 Other #1 Other #2	Switchblade	Striking-Weapon 1 2 Threatening 1 2
Length, Ear 1 2 3			ISM Sebind Inst. I 2 3 U	insetening I Z .
Shoulder 1 2 3	Honorton RIRLRI	SUSP. ACTIONS	Motor Vehicle 1 2 3 U	RESISTANCE BY VICTIM
Below Shoulder 1 2 3	Lowerleg RIRLRL Foot RIRLRL	Unknown 1 2 3 Ate on Prem. 1 2 3	Caustic Chem. 1 2 3 U Poison 1 2 3 U	None 1 2 2
EYES	TOUS REFERE	Unknown 1 2 3 Ate on Prem. 1 2 3 Drank on Prem. 1 2 3	Hot Liquid 1 2 3 U	Before Crime 1 2 During Crime 1 2
Unknown 1 2 3	AMPUTEE	Drank on Prem. 1 2 3	Rope 1 2 3 U	After Crime 1 2 3
Round 1 2 3	Unknown 123 Breast RLRLRL	Chewed Tob./Gum 1 2 3 Intax./Drugs 1 2 3 Smoked on Prem. 1 2 3	Tape 1 2 3 U Wire 1 2 3 U	Uriknown 124
Slanted 1 2 3 Bulging 1 2 3	Entire Arm RLRLRL	Smoked on Prem. 1 2 3	Pass/Hidden Key 1 2 3 U	SUSP./VIC. RELATIONSHIP
Sunken 1 2 3	Lower Arm RLRLRL	Pipe 2 3 Cigar 2 3 3	Bolt Cut./Pliers 1 2 3 U	#1 SUSP. Rel1 = 2 = 3
Squint 1 2 3	Hand RERERE *Fingers RERERE	Cigar Rolled Cig. 2 3	- Orill	Rel 1 2 3
Blink 1 2 3 Crossed 1 2 3	Entiralog ALALAL	Cigarettes 1 2 3	Punch 1 2 3 U	Rel1 2 3
Turned in RLRLRL	LowerLeg RLRLRL	Brand #1	Saw	
Turned Out RLRLRL	Foot REALAL Toes REALAL		Carried Away 1 2 3 U	Unknown Relation 1 2 3
Missing RIRLAL Patch RIRLAL	Art. Limb RLALAL	Disabled Phone 1 2 3	Explosive 1 2 3 U	#2 SUSP.
Auditorial DIGIDI	(Specity)	Left Handed 1 2 3 Right Handed 1 2 3	Flammable Liq. 1 2 3 U	Rel
Blind RIRLRI	TATTOO Unknown 1 2 3	Silent 1 2 3	Torch 1 2 3 U	Rel1 2 2 Rel1 2
EARS	Upper Arm RLRLAL	Talkative 1 2 3	Check 1 2 3 U	Acquaintance 1 2 3
Unknown 1 2 3	Lower Arm RIALRL	Systematic/Org. 1 2 3	Credit Card 1 2 3 U	Stranger 123 Unknown Relation 123
	Hand RLALRL Fingers RLALRL	Used Decoy 1 2 3 Used Driver 1 2 3	Ruse/Con 1 2 3 U	#3 SUSP.
Large 1 2 3	Chest RLRLAL	I Dasg Fookont 5 4 1	Physical Force 1 2 3 U	HeL 1 2 3
Close to Heed 1 2 3	Abdomen RIRIAI	Used Talaphone 1 2 3 1	Threat 1 2 3 U "Special Davice 1 2 3 U	Rel
Protruding 1 2 3 Cauliflower RLRLRL	Back RLRLRL Buttocks RLRLRL	Was Polite 1 2 3 Abusiya Languaga 1 2 3	Other #1	Acquaintance 1 2 3
Pierced RLRLRL	Upperleg RIRLAL	Defecated 1 2 3		Strenger 1 2 3 Unknown Relation 1 2 3
rimited urure		Orinated 1 2 3	Other #3	Unknown Relation 1 2 3
Missing RERERE	Lowerleg RLALRL	, (V.III.	

FIGURE VII-4a INVESTIGATION REPORT CHECK-OFF LIST



1 2 3 6 1 2 3 6	AND EXIT Inknown 1 2 3 3 3 3 3 3 3 3 3	VENDING MACHINE	DOOR	RESIDENCE	C. R. N.
1 2 3 F 1 2 3 F 1 2 3 U 1 2 3		WINDOW By Door 1 2 3 In Door 1 2 3 Crank/Swing 1 2 3 Fixed/Display 1 2 3 Louvered 1 2 3 Skylight 1 2 3 Skylight 1 2 3 Broken 1 2 3 Cut 1 2 3 Tapad 1 2 3 Tapad 1 2 3 Other 1 Other 2 Cut 1 2 3 Cut Cut 1 2 3 Cut Cut 1 2 3 Cut C	LOCK Burned 1 2 3 Cut 1 2 3 Forced Hasp 1 2 3 Picked 1 2 3 Pried 1 2 3 Pulled 1 2 3 Punched 1 2 3 Removed 1 2 3 Removed 1 2 3 Shimmed 1 2 3 In Door 1 2 3 Padlock 1 2 3 Other 1 Other 2 Other 3	VEHICLE Bus 1 2 3 Pessenger 1 2 3 Taxi 1 2 3 Truck 1 2 3 Vending 1 2 3 Other 1 2 3	Partial 1
SAFE/FIREBO Unknown Burned Drilled Explosive Pealed/Pried Punched Removed Unlocked No Force	X 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 3	Other 3 SCREEN 2 3 3 7 18 2 3 7 18 18 18 18 18 18 18	BUSINESS/STORE Chain 1 2 3 Independent 1 2 3 Retail 1 2 3 Wholesale 1 2 3 Norm. Bus. Hrs. 1 / 2 / 2 / 3 3 /	Translation 2 3 3 3 3 3 3 3 3 3	Wetchdog

FIGURE VII-4b INVESTIGATION REPORT CHECK-OFF LIST (Concluded)

Patrol's main objective in responding to a crime scene is to quickly ensure that the victim is cared for and that the offender is described sufficiently quickly and adequately that he might be intercepted Consequently, the responding officer would not be in a mood to run through an extensive list of descriptions to be checked off, which he might deem extraneous anyway, if a fleeing felon were to be quickly apprehended. Our statistical data identify the categories of information found to be most closely associated with case solution. We found little indication that eye, hair, ear, face, or nose characteristics, or mannerisms, had any significance in a post-crime-scene investigation and clearance.

Consequently, we suggest that agencies having extensive computer facilities should seriously reevaluate policies that inflict an enormous paper burden on patrol officers.

Our review of the literature prior to undertaking this project revealed great contrast between "real world" results and controlled laboratory-type experimental findings. We regard the selected studies referenced here as well-designed and well-evaluated experiments. But the conclusions drawn did not take into consideration the operational realities with which we were confronted in analyzing actual crime reports. For example, one study* speculated, on the basis of earlier results by its authors, that:

One might predict that if the number of pictures through which a witness had to search in the identification process could be reduced, the probability of a correct identification might be greatly enhanced. For example, witnesses might provide some preliminary information that would make it possible to eliminate a large number of pictures, keeping only those ... consistent with the verbal description.



K. R. Laughery et al., "Human Memory and the Identification Process," State University of New York at Buffalo (September 1971), p. 33.

A second study concluded:*

Psychological studies of memory for words and pictures reveal that memory for pictures is superior to that for words, and memory for faces is better than that for other pictures. The larger the series of pictures to be recognized the poorer the memory Good identifiers, as opposed to poor, more frequently use facial markings (i.e., unusual features) in identifying suspects. Poor identifiers more frequently use general or intuitive methods in identifying suspects.

Our statistical analyses of the personal appearance descriptors, in particular, reveal an unimpressive low positive correlation with case clearance. Most frequently, we found the correlations with case clearance to be negative. The latter finding indicates that, although the personal appearance descriptors (physical and mannerism characteristics) appear in both cleared and uncleared cases, the overwhelming appearance in uncleared cases causes the correlations to become negative. Thus, the solved cases have been cleared on the basis of investigative information other than physical descriptors. (These factors have been fully documented in Chapter II.)

D. The Investigative Utility of Field Interrogation

One aspect of police investigative operations has not been elaborated on in the earlier chapters: the utility of the somewhat "unofficial" FI or FC (field contact) reporting systems. Many, if not a majority of, agencies use this technique of stopping persons and vehicles for probable cause, even though no crime has been committed. The general rationale for this practice is that stopping persons under suspicious circumstances has frequently led to an arrest for a reported crime, or the persons

^{*}A. Zavala, ed., "Personal Appearance Identification: Psychological
Studies of Human Identification and Recognition Processes," Cornell
Aeronautical Laboratory, Inc. (now CALSPAN Corp.) (Jan. 1970), p. XIII-3.

stopped for interrogation have been found to be in possession of stolen property, dangerous drugs, or narcotics paraphernalia. Although the results of our analyses of the OPD crime reports showed some FC reports associated with case clearances, the actual number of such occurrences was few. Also, we think it likely that the linking of a named suspect to an FC report was an after-the-fact finding.

The OPD does not emphasize an FC program; consequently, the amount of investigatory data in the FC files may not contribute very much to crime-solving leads. Recently, however, a comprehensive evaluation of the San Diego Police Department FI System was published, which reveals interesting if not conclusive findings:*

The analysis supports the hypothesis that some level of FI activity, as opposed to none, provides a deterrent effect on suppressible crimes in localized areas. Further study is recommended to investigate probable area-displacement effects and to identify the factors involved in determining the optimum levels of FI activities. However, there were indications that burglary, petty theft, and malicious mischief/disturbances--crimes most frequently committed by two or more juveniles or young adults--may be the types most influenced.

Taking into account that most (approximately 83%) of the arrests in the Department arise from other than FI activities (such as radio calls), and that more than 98% of field interrogations reported do not result in arrests, it is clear that whatever effects field interrogations have on suppressing crime stem mainly from the FI process itself.

Although the analysis failed to show that the frequency of arrests was significantly influenced by the frequency of Field Interrogations, there were indications that FI activities contributed to 15 percent or more of the total arrests made by patrol officers and that reports of Field Interrogations helped to lead to additional arrests as the result of crime investigation activities.



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J. E. Boydstun et al., "San Diego Field Interrogation Final Report,"
System Development Corporation (Police Foundation, August 1975), pp. 5-6.

The current manual filing and retrieval system employed by the San Diego Police Department effectively prohibits any extensive use of FI reports by investigators. Under these conditions, it appears that the actual utility of FI reports to investigators is minimal, although the potential utility is considered to be high by the investigators themselves. Recently, investigators were provided an improved method (a computer-based system) for comparing FI report data with the suspect information contained in crime reports. The use of this computer-based system is being analyzed.

E. The Necessity for Exposing the Functional Needs of Investigation

In summary, we deduce from the experiential data cited from the literature and from our analyses that the utility of EDP suspect/event-oriented systems is highly dependent on a massive data collection and compilation effort. Furthermore, the success of such systems is critically dependent on the ability of the investigating officers to develop important information that clearly contributes to offender identification. The collection of finely detailed information on a massive scale is not only expensive and time-consuming, but may actually be counterproductive. Therefore, on the basis of OPD operational procedures and results, we conclude: The roles of patrol and detective cannot be viewed as distinct and separate functions.

There should be no mystique about investigative work. The primary requisite is supportive, interactive departmental teamwork to ensure the acquisition of relevant information that will enable efficient sequential case-handling procedures. We view patrol as not only fulfilling a crime-suppressant role but also performing an investigative function. How efficiently the patrol officer documents the events of a crime to which he responds (in which no suspect is apprehended on scene) will have a definite impact on the case outcome as other investigators attempt to pursue the case.

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Our original intent in this project was to write a concluding chapter that would present an "idealized" crime report form that could serve the multiple purposes desired by efficient police case management practitioners and that would also embrace the best offerings of computer technology. But on reflection we now consider that a more constructive approach would be to involve the law enforcement community. We are connected that police agencies themselves are able to devise reporting formats and to design supporting investigative systems to serve their particular requirements. But what is needed for the success of these tasks is consideration of the implication of the facts presented in this report and in the work of others researching the field of criminal investigation and police performance measures.

We therefore recommend that the concluding chapter we had intended to write should be written as an outgrowth of a workshop to be conducted under the aegis of the National Institute of Law Enforcement and Criminal Justice. The issues we have raised and the supportive facts presented here could serve as a stimulus to participating police agencies experiencing concern over the interrelationship between investigative and patrol operations. Furthermore, the contribution of technology as an aid in controlling the criminal population needs to be explored in concert with agencies who have made, and who are contemplating making, heavy financial and personnel resource investments. Too frequently a research report gathers dust on a recipient's bookshelf. But through a workshop drawing attention to important, if not controversial, findings, participating agencies would find more reason to become part of creative policy and decision-making processes that can impact on the growing national crime rate.

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Appendix A

SOCIOECONOMIC PROFILE
OF OAKLAND

APPENDIX A. SOCIOECONOMIC PROFILE OF OAKLAND

The City of Oakland is one of the two major inner-core cities of the San Francisco-Oakland Standard Metropolitan Statistical Area (SMSA). The SMSA contains more than three million inhabitants, which makes it the sixth largest population area in the country. Oakland occupies about 54 square miles, and its 1975 population is estimated at 350,000, which is approximately a 3% decrease from the 1970 U.S. Census data.

Examination of the many neighborhood socioeconomic units that make up the city would provide a more accurate picture of Oakland, but for the purpose of this report, a brief overview is included to give some understanding of the urban unit as a whole. The socioeconomic data examined here pertain to ethnic characteristics, age and sex distribution, and income and employment. Wherever possible, statistical comparisons are made from the 1960 and 1970 U.S. Census data, or more recent information if it is available. These comparisons may provide some indicators as to future trends affecting crime, crime, patterns, and apprehension of law breakers.

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1. Ethnic Composition

Since the 1960 U.S. census, the City of Oakland has shown a trend toward an increasing minority population. This increasing minority population is not inflating the city's total population, since there has been a sceady migration of whites out of Oakland. Table A-1 illustrates the most recent ethnic statistics collected by the Oakland City Planning Department.



Table A-1

CITY OF OAKLAND

ETHNIC COMPOSITION: 1970 AND 1975

	1970	*	1975	Percent Change	
	Population	Percent	Population	Percent	1970-75
Total population	361,561	100.0%	350,000	100.0%	- 3%
White	182,620	50.5	127,100	36.3	-30
Black	122,301	33.8	149,100	42.6	+22
Spanish heritage	35,372	9.8	42,800	12.2	+21
American Indian	2,890	0.8	4,200	1.2	+45
Chinese	11,335	3.1	14,000	4.0	+24
Japanese	2,405	0.7	3,000	0.9	+25
Filipino	3,633	1.0	7,000	2.0	+93
Hawaiian	351	0.1	400	0.1	+14
Korean	222	0.1	2,000	0.6	+801
Other non-white	432	0.1	400	0.1	- 7
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^{*1970} U.S. Census.

In 1969 it was estimated that by 1985 the black population would be a majority of the inhabitants.* Currently, the black population is approximately 43% of the city's total, and the school age population is 65% black. The other minority population segments, except those of Spanish surname, are relatively small.

^{†1975} Oakland City Planning Department estimate, July 1975.

[&]quot;'Options for Oakland: A Summary Report on the Oakland 701 Project,"
City Planning Department, Oakland, California (December 1969).

2. Age and Sex Characteristics

The age distribution characteristics indicate some possibly significant changes that occurred between the 1960 census and the 1970 census. The number of children under the age of 18 decreased by 11.6%. The only major increase was in the 18 to 24 age group of both sexes, which exhibited a 61% increase. A smaller increase (10%) was indicated in the 25 to 34 age group. The population 65 years of age and older remained substantially unchanged, with women outnumbering men by approximately three to two. Overall, the total population of Oakland appears to be becoming younger in that the median age dropped from 35.7 years to 31.9 years.

Even though females exceeded males by 8% in 1970, this represents a drop from a high of 9.5% in 1960. Women exceeded men by 12.5% in the population group over 18 years of age, but the sex distribution for children under 18 was nearly equal. Oakland's work force reflected the national trend of more women entering employment; in 1970 it was 42% female. This was a 3% increase over the 1960 figures. The number of employed females increased by 4,500, while the male employment decreased by 17,000.

3. Income and Employment

Monetary income within the city substantially increased between the two census surveys. The median family income for all economic groups rose 52% during the 10-year span, although no correction for inflation was figured. In 1970, 50% of the employed population earned in excess of \$9,625, and the average salary was \$11,279. A little more than 23% of the families earned more than \$15,000. The percentage of families whose income was below the federal poverty level was 12.2% in 1970, with no comparable data available in the 1960 census; 13.9% of the families in Oakland were receiving some form of public assistance in 1970. Unemployment continues to be a major problem for the city. The July 1975 labor



statistics indicate that Oakland experienced a 13.9% unemployment rate as compared to 9.9% for the Bay Area and 8.4% for the nation as a whole.

White-collar jobs significantly increased by 21% between 1960 and 1970. As a result of this increase, 57% of all the jobs available in the city were white-collar in 1970. Blue-collar jobs decreased by 10% during the same period. Service workers had increased by 21%, although they represented only 15% of the total labor force in 1970. Manufacturing jobs had declined, and this trend was projected to continue. Oakland's labor force appears to be in a state of transition, because the city is becoming one of white-collar service workers, technicians, and professionals.

Appendix B

SURVEY OF THE OPD AND ITS INVESTIGATIVE PROCEDURES

APPENDIX B. SURVEY OF THE OPD AND ITS INVESTIGATIVE PROCEDURES

1. OPD Organization and Crime Reporting Procedure

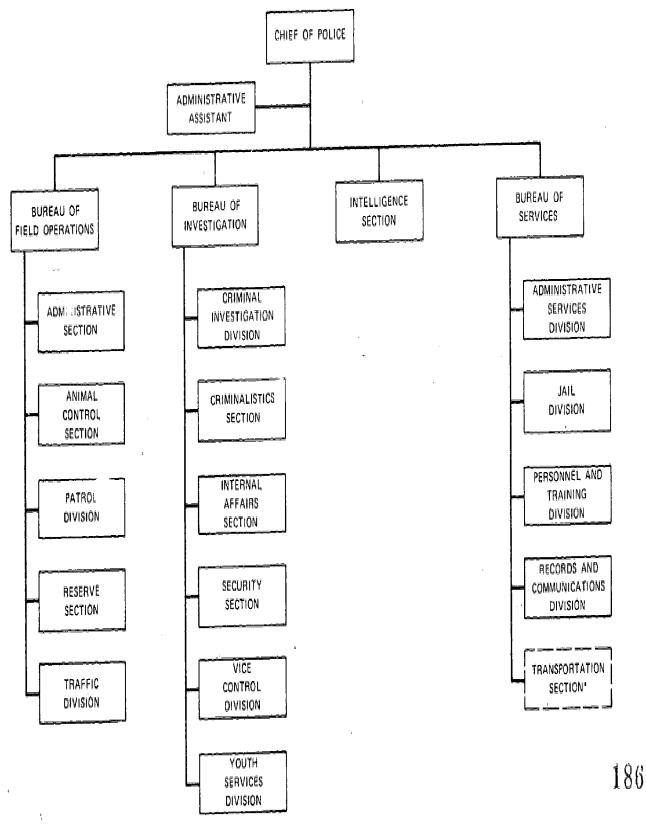
At the initial stage of the project, interviews were held with the OPD Chief and Deputy Chief designated to assist the effort. Introductions were made to appropriate Division Commanders and investigative and records surpervisory personnel. Tours were conducted of the divisions containing information of potential value to the project's objectives. Given below is a summary of how the OPD is organized. Documents we analyzed that were related to reports of felony crime investigations are illustrated.

As of 1973, the OPD was authorized the following personnel: sworn, 722; civilian, 280; and 76 auxiliary officers. The Department is organized into three bureaus, each commanded by a Deputy Chief, and one section. Figure B-1 shows the three bureaus and the one section reporting to the Chief: Bureau of Field Operations; Bureau of Investigation; Intelligence Section; and Bureau of Services. Our research activities were largely confined to the Criminal Investigation Division (CID), Youth Services Division, Records and Communications Division, and Patrol Division.

The CID is commanded by a captain. Six lieutenants supervise special sections, such as Auto Theft, Burglary/Theft I and II, Homicide, Robbery, and Forgery/Fraud. Currently, 72 sergeants, 5 inspectors, 12 police officers, and one policewoman make up the CID sworn personnel complement.

During the period from which our sample was drawn, the city was organized into 29 beats. As of January 1975, however, the OPD realigned its patrol operations after considerable study and became organized into 35 beats under what is known as the Patrol 35 plan. The major objectives





185) *Functional Supervision of the Equipment Department's Transportation Section.

FIGURE B-1 ORGANIZATION CHART-OAKLAND POLICE DEPARTMENT



of the reorganization were to improve patrol response to citizens' calls and to improve the overall delivery of police services. Selective riding with patrol by SRI staff occurred after the new beat structure had been implemented, but we made no attempt to determine whether the revised operations had had any impact on crime suppression or clearance levels.

Briefly, the flow of crime incident information generally originates by a call to OPD headquarters and is handled sequentially by any of several operators in the communications and dispatch section. The OPD operator evaluates the importance of the complaint and records the critical information on the Complaint-Dispatch form (Figure B-2). If a crime is in progress, the operator time-stamps a red-border card and places it on a conveyor for immediate handling by the dispatchers. The patrol vehicle status board is scanned for an available beat unit, and dispatch orders are given, together with as much information as is needed to inform the responding unit(s) of the nature of the crime and the offender's description. According to the new Patrol 35 Plan, dispatch is given the responsibility to control vehicle deployment, and the card is filed in the slot designated for the assigned unit. In practice, however, multiple units may respond, depending upon how the adjacent beat officers view the seriousness of the crime and their freedom or desire to assist. A nonpriority complaint dispatch card is handled on a unit-availability basis.

Located in the communications room is a direct-wire annunciator panel of silent hold-up and intrusion alarms linked to important facilities such as banks. In response to an alarm, immediate dispatch of patrol elements is made, and a call to the installation is also made. Specific information is requested regarding a crime in progress. Descriptions of the perpetrators are obtained and put on the air. The OPD has been fairly successful in apprehending bank robbers by such quick and efficient procedures.

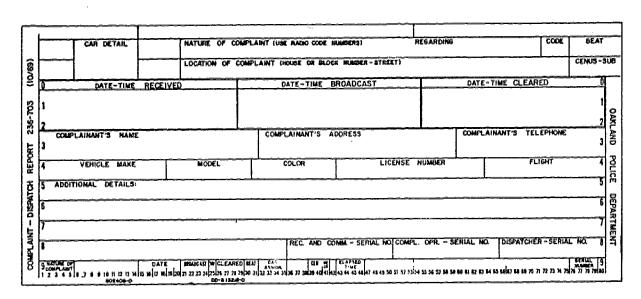


FIGURE B-2a OPD COMPLAINT DISPATCH REPORT (OBVERSE)

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			GLASSES		MUSTACHE	GLASSE3		MUSTACHE
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FIGURE B-2b OPD COMPLAINT DISPATCH REPORT (REVERSE)



All complaint-dispatch cards are filed away in boxes after the responding unit has indicated that the assignment has been completed and the unit is back in operation. No serial numbers are assigned to these cards, and no provision is made to link the cards to specific incident reports for possible use by investigators. Whether any potential information might be available or lost through the medium of the complaint-dispatch form may be worthy of some analysis. The Richmond, California, Police Department evidently feels the complaint-dispatch form is of value, for a case number is immediately assigned to it. It is then possible to bring the report of incident and the dispatch card together for a complete record of the event.

When there is a crime incident to which a patrol officer responds (and supporting units, e.g., special operations section, detectives, or helicopter), an official report is prepared. The basic form used is the Crime Report (Figure B-3). This report form is prepared in longhand and filed at Headquarters. It is then immediately processed for EDP recording of certain information. Tapes and printouts on incidents are created daily by the City of Oakland Data Processing Center, for statistical and operational uses. The OPD Research and Development Section controls this procedure. We found that, although the computer tapes and printouts were useful to identify all the crime incidents and certain data as to locations and times by report number, the effort to scan manually for felony offenses by felony category (penal codes) would have been too great for our purposes. Consequently we reprogrammed copies of the OPD tapes to facilitate the indexing of felony reports of interest.

Basically, the general OPD crime report, compared to others we have worked with, is well-designed. If there is a need for more descriptive data on an incident, supplementary information can be continued on the Additional Information Report (Figure B-4). Incidents regarding vehicles are described in the Vehicle Report (Figure B-5).



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FIGURE B-3 OPD CRIME REPORT



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FIGURE 8-4 OPD ADDITIONAL INFORMATION REPORT



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· FIGURE B-5 OPD VEHICLE REPORT

The OPD has a limited field contact procedure. The form is illustrated in Figure B-6. If an individual is stopped by an officer on probable cause when no crime has been committed, a field contact (FC) card is prepared if the officer feels there may be "suspicious circumstances" surrounding the person or vehicle. Two procedures are then followed. If a vehicle is involved, the information is processed for entry into the automated Crime File system. If only people are involved, the cards are filed alphabetically by year in a cabinet. The Crime Analysis Section, described later, conducts searches for FC reports when a suspect has been named but not necessarily apprehended.

If the responding officer decides that the crime scene (for specific priority felonies) warrants the services of an evidence technician to search for and recover relevant physical evidence, he makes a request. The technician compiles the Technicians Report (Figure B-7). Processing of physical evidence, particularly fingerprints, is undertaken by the Criminalistics Section. Its report is filed on the Criminalistics Section Service Request form (Figure B-8).

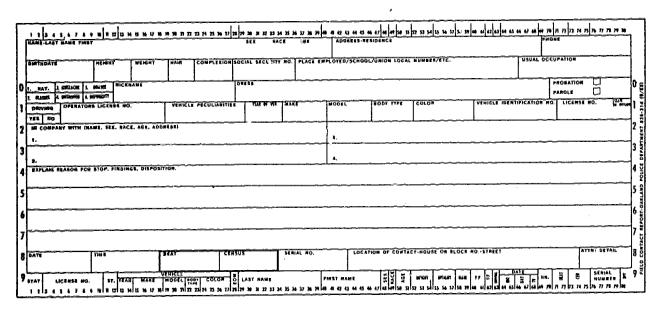


FIGURE B-6 OPD FIELD CONTACT REPORT



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ARKARA	EATE AND TIME CRIME OCCURATED	LOGATION OF OCCURRENCE		LAB. NO.
			. 1	

FIGURE B-7 OPD TECHNICIAN'S REPORT



CRIMINALISTICS SECTION SERVICE REQUEST	R 3. NO.
Oakland Police Department	LCF NO.
NAME OF VICTIM OR COMPLAINANT ADDR	
TECHNICIAN'S REPORT ON FILE? YES NO	
CRIME DATE OF CRIME LOCA	TION
RELATIVE URGENCY OF ANALYSIS REQUESTED	
SUSPECT IN CUSTODY CAN HOLD CANNOT HOLD WITHOUT RESULTS	G OF ANALYSIS.
Account	
SUSPECT NOT IN CUSTODY - NOT URGENT.	
WILL BE NEEDED FOR PRELIMINARY HEARING.	
NOT NEEDED UNTIL SUPERIOR COURT TRIAL.	
REQUEST IS MADE FOR:	· · · · · · · · · · · · · · · · · · ·
FINGERPRINT COMPARISON	
OTHER TYPE OF ANALYSIS - BE SPECIFIC - PROVIDE PROPERTY PAGE AND ITEM	NUMBERS OF EVIDENCE YOU WANT
OTHER TYPE OF ANALYSIS - BE SPECIFIC - PROVIDE PROPERTY PAGE AND THEM	Company with the second
SUSPECTS OPD NO.	
14	
2 5	
3 6,	
REQUESTED BY	TELEPHONE EXT.
APPROVED BY DATE	
LABORATORY EXAMINATION COMPLETED BY	DATE
1	
SUMMARY OF RESULTS OF LABORATORY EXAMINATION	
,	
,	
	•
•	
MAKE 3 COPIES. FORWARD ORIGINAL AND ONE COPY TO CRIMINALISTICS SECTION.	

FIGURE B-8 OPD CRIMINALISTICS SECTION SERVICE REQUEST

149

195



884-269 (7/71)

Depending on the seriousness of the crime, a detective may be active in the initial investigation. Generally, however, the CID investigators respond after the initial crime reports, and supporting documents are forwarded to them. A record of the assigned CID investigators working on a case is prepared on the Follow-Up Investigation Report [Figure B-9(a)]. The unique feature of this form is Item 16, which indicates the official disposition of the case. The OPD policy is that only the investigator can officially assign a disposition to a given case. The reverse side of the form [Figure B-9(b)] is an Investigator's Check-Off List to remind the investigator to pursue the procedures specified. We rarely found that check-off entries had been made on the reports we analyzed in the CID record files. CID policy is that, except for homicide, a 30-day suspense time is imposed for completing an investigation. Reasons for delays must be cited.

Figure B-10 shows the form used by the interviewing officers when taking statements. We frequently found that information contained in this form was not contained in the crime report or the supplementary forms. This information has a certain value regarding descriptions of events and persons involved.

Alameda County has developed a Consolidated Arrest Report (Figure B-11) which is used by all jurisdictions in the county. Certain key information in the uniform arrest report is transferred to the County CORPUS EDP system (Criminal Oriented Records, Productions, Unified System). CORPUS terminals are provided at the OPD and the county jails so that a PFN (person filing number) for booking can be assigned permanently to the arrestee. If the subject has had a prior arrest and conviction record, the same PFN number is recorded on the new arrest report (bearing the assigned case number). A CORPUS terminal is also located in the CID to assist in suspect identification, and terminals are presumably becoming available in other Alameda County law enforcement agencies.

FOLLOW-UP INVESTIGATION R OAKLAND POLICE DEPARTM		1. DATE OF	THIS REPORT	2. RO NUMBER		3. FOLLOW-UP REPORT NO.
OAKLAND FOLICE DEFARTA	5. CH	ANGE COMPL	AINANT'S NAME	AND ADDRESS TO		
			Euros come	AND CLASSIFICAT	TION TO	
ORIGINAL CRIME AND CLASSIFICATION		/′	CHANGE CHIME	AND CEASSIFICA	TON TO	
, TYPE OF PREMISES	9, OBJEC	T OF ATTACH	(IO. DATE OF	ORIGINAL REPORT
1, WHERE AUTO WAS RECOVERED	<u> </u>	1â	. VALUE OF LOS	99		ECOVERED PROPERTY
4. NAME OF SUSPECTS RESPONSIBLE	SEX:	RACE-DOB	AP	REST NUMBER	OPD NU	MBER
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DATE 15. ENTER THE RESULTS OF EAC	H STEP OF YOUR	INVESTIGAT	, I ØK			
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☐ 04 Pros. for Another Offense☐ 12 D.A. Citation		d Property	urned Home Ret. to Own 'S NAME SERIA	er 🗍 16	Notice to Ap	pear
OX Unfounded OR Pros. By Outside Department				1		

336-201 (REV. 5-71)

FIGURE B-9a OPD FOLLOW-UP INVESTIGATION REPORT (OBVERSE)

INVESTIGATOR'S CHECK-OFF LIST OAKLAND POLICE DEPARTMENT

OARLAND POLICE DEPARTMENT	YES	NO	NOT APPLICABLE
MO CHECK	, , , , , , , , , , , , , , , , , , , ,		
VISIT CRIME SCENE	,		
CONTACT COMPLAINANT			
CONTACT WITNESS(ES)			
INTERROGATE SUSPECT(S)			
LINE-UP ON SUSPECT(S)			
ROGUES GALLERY RUN			
INFORMANTS			
REYIEW TECHNICIAN REPORTS			
CHECK WITH CRIMINALIST (PRINTS, ETC.)			
EYIDENCE REVIEWED			
EVIDENCE DISPOSED OF			
VEHICLE RELEASED			
APB			
DAILY BULLETIN NOTICE			
YOUTH SECTION CONTACT			
OUTSIDE AGENCY CONTACT:			
Parole			
Probation			
Other Departments			
CRIME ANALYSIS SECTION:			
Field Contacts			
Yehicle Listings			
Known Offenders			
Recap Information			
CONFERENCES WITH PROSECUTING ATTORNEY			
CASE REPORT MADE			

INVESTIGATOR	BERIAL NO.	APPROVING SUPERVISOR	BERIAL NO.
		•	444

FIGURE- B-96 OPD FOLLOW-UP INVESTIGATION REPORT (REVERSE)



1. NAME OF COMPLAINANT OR DEFENDANT		2. RE	PORT NUMBER			TATEMENT				
3. NAME OF PERSON GIVING STATEMENT	SEX RACE	DOB 4. RE	SIDENCE ADDRESS		Oakland Police I	ONE				
5. EMPLOYMENT INAME · ADDRESS · PHONE - OCCUP	ATION) OR SUPPLEM	ENTAL INECHA	ATION IE LINEARI OVER	OR TRANSIE	N T					
			The state of the s	on manage						
8. STATEMENT TAKEN BY.	<u> </u>			f no						
name	30/10	il no.	date	ti	me started time	completed				
ATIN PRESENCE OF										
location where statement taken names and addresses of persons present when statement taken FOR VEHICLE COLLISIONS ONLY										
VEHICLE YEAR MAKE MODEL	TYPE CC	COLORISI LICENSE NO. IOUT ONLY			OPERATOR'S NO.	STATE (Out Only)				
REGISTERED OWNER	AC	DORESS		CITY	RES. OR BUS. PHO	INIE				
ADMONITION: YOU HAVE THE RIGHT TO REMAIN SI RIGHT TO TALK TO A LAWYER AND HAVE HIM PRES APPOINTED TO REPRESENT YOU BEFORE ANY QUEST	ENT WITH YOU WHILE	E AUTH THE BEIN	ID WILL BE USED AGAI IG QUESTIONED, IF YO	NST YOU IN U CANNOT A	A COURT OF LAW, YOU HAFFORD A LAWYER ONE W	IAVE THE				
WAIVER:	IONING. IF YOU WIS	SH ONE.								
DO YOU UNDERSTAND EACH OF THESE RIGHTS I H.	AVE EXPLAINED TO	YOU?								
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FIGURE B-10 OPD STATEMENT



ALAMEDA CO									-		1		lee.			Agen	y: CA	ortmen 0010
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4 True Name						g	Date o	f Arres	τ	6 So	cial Sa	curity No.						
7 AKA or Nickna	rhe					8	Time	of Arres	it .	9 Mi	scellan	eous I,D, No.	1					
10 Address					С	ity		s	tete	11 P	hone		Finger					
12 Employer					·· —·~	<u>-</u> -	Pho	 ne		13 C	Ccupa	llon	Index F					
14 Sex Race	Hair	Eyes	Height	We	eight	Dat	e of Bir	th ,	Age	16 S	upervis	ing Sgt,	Right In					
1 1 18 POB Bai	d	Skin	Glas I	1 Mar	Det	L	cz	- L YA	1CO	70	CAL	YADR		Numbe	r		Com	plete
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() 30 Arresting Office		[]	[] []	lficer		No		2 Tran	NEDOSTI	na Off	ficer	No.	Cour	t Appe	arance	,	a.	a.
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33 Hold For					3	4 Ajutho	ority	35	Hold	Relea	ted		Coli	t Date	& Tim	•	Signature	Signal
36 Vah, License No	, Stat	e Year	37 Yaar Mak	0	M	odel	C	olor	38	Towed	To			Hold ()	39	Evid	ence H	eld [] No
40 I hereby arrest t			n the charge indic	ated an	d reques	T A POAC	e officer	to take	e him	into c	ybofsu	I will appear	es dire		nd sign			
										Signed								
Name		····	· 	Add	ress		10.0				Home		44 5		isiness			
41 Person to Be No	tifind		[] vut	Eme	re []		42 H	esidence	e Phor	۱ ۱		sistionship	<u> </u>		me of	NOTITE	Cation	
46 Address of Perso	on to b	e Notified					46 Bu	usiness f	Phone		47 01	ficer Who Ma	de No	ificatio	on			
48 Co-Defendant		Sex	- Asce - Age	Co-D	efendan	ŧ		S	6x f	Race -	Age	Co-Defenda	nt			Sex .	- Race	- Age
19 Completnent's N	lame			L		.Adi	ires.	·			,	Cit	ν .			<u></u> .	State	·
O NARRATIVE II	VSTR	UCTIONS:	A. List all charges of other arrests	by nam	ne and co	de secti	on: B. It	temize n	multip	ie che	rges; C.	Document V	our ad	monish	ment :	of the	orreste the total	d per-
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52 Prisoner Signatu		tgoing (53	Officer	Signatu	re/Qutg	oing			Date &	Time	54 R	eaton	for Re	gate		
56 Juv. () Int			[] Other Agency		56	() Pin			57		С, Н,	1 7 1	COSS	59) []	To	Ċ	kd, By:
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FIGURE B-11 ALAMEDA COUNTY CONSOLIDATED ARREST REPORT (OPD)



A useful instrument, specifically designed for use by the Alameda County Office of the District Attorney, is the Prosecutor's Case Summary (Figure B-12). Recognizing the case loads carried by prosecutors and the voluminous records (as shown previously) that can be amassed on suspects arrested for alleged crimes, OPD case investigators prepare a brief of a case to facilitate a complaint to be signed by a given deputy district attorney. A complaint issued by the prosecutor, in effect, provides the basis for the OPD case disposition category "O1-Arrest and Prosecution" shown in Figure B-9(a) (Follow-up Investigation Report).

2. Crime Analysis Section

We found it significant that the OPD had set up a Crime Analysis (CAS) unit to review all incident reports to determine whether additional information could be provided to assist investigators in following up certain cases. The intent was to minimize the number of detectives needed to undertake routine tasks. Consequently, we made a special effort to evaluate the effectiveness of the CAS report enrichment procedures insufar as their efforts contributed to case solutions. Unfortunately, the CAS system had not been formalized and running, as described below, for a sizeable portion of the sample felony cases we drew for July, August, and September 1974.

The CAS procedure is as follows: When patrol turns in the incident reports prepared, they are sent to the Report Reproducing Unit (located in the CID). The reports are assigned a number, and a computer card is prepared for complainant reference and for preparation of statistical reports. After reproduction of the original report in multiple copies, CAS staff receives the report for staff review. Each report is coded by an alphanumeric designation indicating the priority handling (letters A-J) and type of information search to be conducted (numbers 1-15).

OAKLAND POLICE DEPARTMENT

PROSECUTOR'S CASE SUMMARY

DEFENDANT'S MAME	SEX RA	ACE AGE	DOB	CHARGE		CASE NUMBER
DEFENDANT'S HES. ADDRESS AND TE	LEPHONE				WHERE OR BY WHOM	EMPLOYED
DATE AND TIME ARRESTED				LOCATION OF ARREST		ARREST NUMBER
ARRESTING OFFICER	SERIAL NO.	ARRESTING OFF	ICER	SERIAL NO.	INVESTIGATING OFFIC	ER SERIAL NO.
VIGTIM'S NAME	SEX R	ACE AGE	DOB	RES. ADDRESS AND PHONE	BUS	INESS ADDRESS AND PHONE
DATE AND TIME OF OFFENSE				LOCATION OF OFFENSE		
THE FOLLOWING DATA SH OF OFFENSE. (3) SUMMARY OF EXI ABLE CAUSE, WARRANT, ETC. (6) C SION, DENIAL, ADMISSION, REFUSE ADDITIONAL DEFENDANTS - ENTER S	PECTED TESTIMONY O DEFENDANT'S CRIMINA D. (8) OTHER REMAR	P WITNESSES. (- AL RECORD (ATT/ KS. (9) SIGN A	4) PHYSICAL ACH RAP SHI IND DATE R	EVIDENCE AND WHERE PRESE EET), (7) IF STATEMENT TAKE	NTLY LOCATED, (5) G	ROUNDS FOR ARREST, I.E. REASON:
WITNESS' NAME		AGE AGE	RES. A	DORESS AND PHONE	BUSINE	SS ADDRESS AND PHONE
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l.						
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055 336-302(11:68)

FIGURE B-12 OPD PROSECUTOR'S CASE SUMMARY





The priority case handling letter code designations are summarized as follows:

- A. Investigation of offenses involving great bodily injury and other major or serious crimes against the person.
- B. In-custody felony suspects.
- C. Named felony suspects not in custody posing a major threat to society.
- D. Major property loss.
- E. In-custody misdemeanor suspects.
- F. Named felony suspects not in custody and not posing an immediate threat to society.
- G. Other felony offenses.
- H. Misdemeanor offenses in which losses are above a given level.
- Misdemeanor offenses in which losses are below a given level.
- J. Investigation of violation of local regulatory ordinances.

Figure B-13 is a copy of the CID Investigators Information Sheet.

The arabic numbers represent sources of information, indicated by a CAS staff analyst, to be checked. Thus, an ADW report having a named suspect not in custody could be coded as C-2,-3,-5,-14. The Data Collection Form, Card 7 (see Appendix C), is the source from which we could ascertain whether the information searches requested and conducted by the CAS staff had produced "useful leads" and/or had linked stolen pro__ty, vehicles, or firearms to a suspect. Computer hard-copy printouts, FC reports, and rap sheets are attached to the incident report if useful information has been obtained. In the case of Item 3, Crime File run-person, the computer operator interrogates the data base by inputting subject descriptors. (The computer system is discussed in Chapter VII.) The TT terminal prints the number of hits, that is the possible suspects who match the descriptors.





CRIMINAL INVESTIGATION DIVISION INVESTIGATIVE INFORMATION SHEET

		CRIME:
Complainant:	i ali sa sain a sura sa	RD#:
1 1028 Registration	8	Stolen Article Run
2 Warrant Information	9	Teletype (property loss) to DOJ
3 Crime File Run - Person	10	Pin Map Data
4 Crime File - Vehicle	11,	Firearms Query
5 Field Contact Information	12,	Corpus Information
6 Crime Re-cap Logs	13	Vehicles Registered to Suspect
7 Driver's License Physical Data	14	Firearms Registered to Suspect
	15	Other
	· 	
Crime Analysis Section personnel will provif applicable.	ide the above	listed information to investigators
REMARKS:		

FIGURE B-13 OPD CID INVESTIGATIVE INFORMATION SHEET

If the number is small, say five or ten, the operator gets a printout of names. With any larger hit range, it is left to the investigator to determine whether to pursue further checking. A companion CRT display is associated with the Crime File System. After the TT terminal has printed the names of suspects whose descriptions match the input query, mugs can be called up for review by witnesses. The relative success of the CAS report enrichment process has been discussed in Chapters II and VII, where the specific variables (such as Crime File run--person) has been ranked according to their contribution to the felony category clearances we analyzed.

The CAS has access to data and sources as shown in Table B-1.

In addition to the Crime File computerized system for known offender and vehicle descriptions, a consolidated CRT display and hard copy printer terminal in the CAS provide centralized access to the county, state, and federal investigative data banks.

In August 1974, the CID issued a directive charging the CAS members with the responsibility of forwarding crime reports to appropriate sections when there was sufficient investigative information to permit a follow-up investigation.

The OPD had been informed, by an OPD research group, of the earlier work published by SRI, on a burglary case follow-up decision model.

Although it appears that, initially, the concept was adapted to OPD needs, the culling of reports by the CAS has evidently been abandoned. The CID detectives read all reports.

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Table B-1

OPD COMPUTERIZED AND MANUAL INVESTIGATIVE DATA SOURCES

Data	Data Bank and Access
Warrant information	Alameda County (and Bay Area) Police Information NetworkPIN
ter .	California Department of Justice Criminal Justice Information SystemCJIS
	California Law Enforcement Telecommunications SystemCLETS
	National Crime Information CenterNCIC
	Wanted Persons SystemWPS
Known offenders' descriptions	OPD Crime File System
	CORPUS
	CJIS
	NCIC
Driver's license physical data	Department of Motor Vehicles, Automated Name Index DMV/ANI (CLETS)
Field contact information	OPD Crime File System
	OPD manual file
Crime incidents	OPD crime recap manual logs
Vehicle information	PIN; Crime File System; NCIC; DMV/ANI; Stolen Vehicle SystemSVS (CLETS)
Stolen property information	Automated Property System (California Department of Justice) APS (CLETS)
	NCIC
Firearms information	CJIS Automated Firearms System (AFS)(CLETS)



3. Other CID Investigative Resources

For cases involving juveniles, the Youth Services Division (YSD) handles the follow-up investigations, and all records on the juveniles are retained by the YSD.

Adult criminal histories are retained in the CAS area on microfiche which is obtained from the Central Identification Bureau (CIB) run by the Alameda County Sheriff's Office. Rap sheets are requested from the California Department of Justice to supplement the county criminal history records.

The OPD has a Criminalistics Section which processes physical evidence. This section also has a CRT display of fingerprints filed by a digital code. This is a subsystem of the CID Crime File System. If good quality latent prints are brought in, the criminalists classify them and interrogate the computer. Comparison is then made if the hit range is of reasonable size. No special effort was made to tour this operation, because we were familiar with its operation. However, the contribution to case clearance by means of physical evidence was analyzed in context with each felony category described in Chapters II through V. The presence of a Technicians Report (Figure B-7) or a Crime Lab Report (Figure B-8) in a given case report provided us with the information as to whether physical evidence had been useful in the investigation.

Because the current case investigation files are more complete in the CID, we elected to work from this source. The Records Division keeps hard copies of reports for two years. In some instances, especially for rape cases, we reviewed the Records Division files, because the Rape Investigation unit and its files were not housed in the Police Administration building. The Records Division also maintains a daily updated alphabetical microfilm listing of names (complainants, victims, and arrestees) associated with certain event information and the report numbers. This "com-alpha" index system was useful in tracking some suspect names.



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The Research and Development (R&D) Section of the Administrative Services Division is the unit that compiles statistical data on crimes. At first, we used their system of indexing reports to identify the report numbers of felonies that we wanted to sample. Using this printout proved much too time-consuming and cumbersome for our need to draw a large-scale sample of cases. Consequently, the R&D Section procured copies of computer tapes of reports, which we reprogrammed into a more convenient format for our purposes.

The OPD R&D Section also undertakes new programs, such as the design and procurement of the now operational Crime File System.

5 DE

Appendix C

DATA COLLECTION FORMS

FELONY INVESTIGATION DATA COLLECTION FORM -- PROJECT #3674 FIRST CARD -- FELONY CLASSIFICATION

Card type $\frac{1}{(1)}$ $\frac{1}{(2)}$	Primary Felony Offen (see below)	R.D.) (8) (9) (1 ght justify)	
Data Recorder's Initials (12	(13)(14)	Date Compiled			
Location of Incident:	$\frac{\text{Beat}}{(19)(20)}$	Census (21) (22) (23)	Date of Occurrent (month/day)	ce (<u>24)</u> (<u>25)</u> (2	26) (27)
Time of Occurrence (2400 %r) (28) (29) (29)	Range of time (2400 hr)	: Between $(32)(33)(34)(35)$ an	d (36)(37)(38)(39)	Day of '#€	eek (40)
Date Reported (month/day) (41) (42) (43) (44)		ted (45) (46) (47) (48)		1 = Mon 2 = Tu 3 = Wed 4 = Th	5 = Fri 6 = Sat 7 = Sun
Additional Offenses Charged	(49) (50) / (51) (52) / (53) (5	/ (55) (56)	! ,	4 - 111	
Case Disposition $(57)(58)$	Date (59) (60) (61) (62)		Date (64) (65) (66)		
Case Uncleared (68)	Closed by Admission (Complaint Signe	d/Date (70)(71)(72)(73	<u>, </u>	
Warrant Issued (74)	Date Released to CID (75)(Priority H	andling Code (79)		
Offense Designation	Offense Code EDP	Offense Designation	n Of	fensa Code	EDP
Strongarm robbery Armed robbery	() = 1 () = 2	Theft: From person	,	1	= 9
Felony assault	() = 3	Pursesnatch	(ý	= 10
Burglary	() = 4	Shoplifting	(,	= 11
Auto theft	() = 5 () = 6	Other (state)		,	= 12 = 13
Homicide, willful Forcible rape	() = 0 () ≈ 7	Narcotics and drug Stolen property	s (,	= 13 = 14
Attempted rape	() = 8	Vehicle laws	(= 15
•		Other	()	≈ 16
		(state)		

 $^{^{\}star}$ Code from bottom of Follow-up Investigation Report.



Card type $\frac{2}{(1)}$	Primary Felony Offense ${(3)}$	R.D. Report No.	(7) (0) (10) (11)
\-/ \-/	(Name and Rank)	(ID Numbers)	(7) (8) (9) (10) (11) Last Date Case Handled:
Reporting Officer(s)	·	$(\overline{12})(\overline{13})(\overline{14})(\overline{15})(\overline{16})(\overline{17})$	(18) (19) (20) (21)
, <u></u>		$(\overline{22})(\overline{23})(\overline{24})(\overline{25})(\overline{26})(\overline{27})$	$(\overline{28})(\overline{29})(\overline{30})(\overline{31})$
Investigating Officer(s)	v.'	(32) (33) (34) (35) (36) (37)	(38) (39) (40) (41)
		(42) (43) (44) (45) (46) (47)	(48) (49) (50) (51)
Arresting Officer(s)		(52) (53) (54) (55) (56) (57)	
		$(\overline{58})(\overline{59})(\overline{60})(\overline{61})(\overline{62})(\overline{63})$	
Evidence Technician	· · · · · · · · · · · · · · · · · · ·	(64) (65) (66) (67) (68) (69)	

Date at Scene (month/day) $(\overline{70})(\overline{71})(\overline{72})(\overline{73})$

Time at Scene (2400 hr) (74)(75)(76)(77)

٦,

Facility Category (see list #2) (21)(22)

List # 1 Location:	List # 2 == Facility Category	
Street_	Residential	Commercial (cont)
0 = Not stated	0 = Not stated	18 = Bank
l = Residential area	l = Apartment	19 = Gas station
2 = Business district	2 = Hotel	20 = Phone booth
3 = Sidewalk	3 = Motel	21 = Other
4 = Parking lot	4 = Single family	(state)
5 = Isolated	5 = Multi*family	nuklia
	6 = Other	Public
Recreational	(state)	22 = Not stated
6 = In building	Campanada 1	23 = School
7 = Open area	<u>Commercial</u>	24 = House of Worship
8 = Wooded or shrubbery area	7 = Not stated	25 = Place public assembly
9 = Rest room	8 = Restaurant	26 = Other
n21.12	9 = Bar	(state)
Building	<pre>10 = Food store/supermarket</pre>	Transportation
10 = In premises	11 = Liquor	¥
11 = Ground floor	12 = Industrial mfg.	27 = Bart
12 = Upper floor	13 = Retail, large	28 = Bus
13 = Elevator	14 = Retail, small	29 = Taxi
14 = Grounds	<pre>15 = Business office</pre>	30 = Auto
15 = Hallway	<pre>16 = Medical office</pre>	31 = Other
16 = Doorway	17 = Pharmacy	(state)
17 = Other		
(state)		

Primary Felony Offense $\frac{}{(3)}$ R.D. Report No. $\frac{}{(5)(6)(7)(8)(9)(10)(11)}$ Citizen Informant Police Informant Victim ____ Witness Reported By: (14)On View No. of Reporting Individuals Alarm ___ Principal Reporting Individual(s) Providing Useful Information: a) Age group Sex Race (20) (see list below) c) Age group Sex Race (27) b) Age group (22)Sex Race (24) Hospitalized ____ Uninjured (28) Minor, not Hospitalized Minor, not Hospitalized (29) Victim's Condition: not serious (30) serious (31)Dead (32) Cooperative Lucid (Y/N) (33) (Y/N) $(\overline{34})$

Weather and Illumination Conditions: Rain / Fog / Clear / Unknown / (record only if stated in report) (35) (36) (37) (38)

Daylight / Dawn / Dusk / Dark / Artificial (B = Bright, D = Dim)

	Age Group	Sex	Race	
· !	A = Adult J = Juvenile	M = Male F = Female Ø = Other (state)	W = White N = Negro M = Mexican I = American Indian	C = Chinese J = Japanese Ø = Other (state)
		×	Salant,	

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Primary Felony Offense (3) (4)

R.D. Report No. (5) (6) (7) (8) (9) (10) (11)

Weapon used/ Physical force used Description provided of Special attempts to conceal identity observed (12)(13) & injury inflicted (14) physical attack mode (15) by stealth or handling of victim (16)

(list #1)

Sexual actions and/or Vehicle used Description: make Color License No. (21) or taken (18) model, type, year (19)

(21)

2 = Other see Crd. 7

Property Taken: Cash, Value / Clothing= Value / Jewelry, precious Value / (see list #2) Negotiables (22) (23) furs (24) (25) metals (26) (27)

Firearms Value / Office Value / TV-Radio Value / Wousehold Value / Wousehold (35)

 Consumable
 Value
 / Livestock
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 / Hardware
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 / Stocks Value
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Liquor Value / Auto / Auto theft- / Credit cards / Wallet / Purse / (44) (45) theft (46) stripped (47) (48) (48)

Checks / Drugs Value / Other Value (53) (53) (54) (55)

(state)

/ Total cash value

Property Identification Indicated:

Description ___ (57)

Serial no.(s)

List # 2 Cash Value

List # 1 Weapons

1 = Handgun

6 = Explosive

2 = Rifle

7 = Chemical

3 = Shotgun

8 = Blunt instrument

4 = Alleged gun 5 = Knife

9 = Simulated

10 = Other ___

11 = Unknown

(state)

6 ≈ \$2,000 - \$5,000 7 **≥** > \$5,000

5 = \$1,000 - \$2,000

3 = \$200 - \$5004 = \$500 - \$1,000

2 = \$100 - \$200

1 ≈ < \$ 100

Card type $\frac{6}{(1)}$	Primary Felony Offense	(3) (4)	.D. Report No. (5) (6) (7) (8) (9) (10)	<u>(11)</u>
Classification of Entry: S	treet crime Forc	ed entry Uni	lawful entry (14)	Invited in (15)	No Indication (16)
Mode of Entry: Door (17)	Window Roof (19)	Wall Bar (20)	Sement Conce	nied Other $(\overline{22})$	23) (state)
Object of Attack: Proper	rty <u>Pers</u> (24)	ona <u>(25)</u>	Safe (26)		
Physical Evidence Cited and/or Recovered and Matched from a) Crime Scene B) Suspects	r a) Toolmarks b) Tool / a) To match (28)	ools b) Match (39)	/ a) Fingerprin	nts- b) Match /
	a) Footprints (33)	Shoes / a) 1 $(\widetilde{34})$	Piremarka b) Ti: (35)	res / a) Weap (36)	ons (37) b) Match (38)
a) Clothing b) Match (40)	/ a) Bloodstains (41)	o) Match / a) T (42)	race Materials (43)	b) Match	(state)
a) Other Physical evidence (45)	b) Match	(state)		:	
Indication of alarm or telepho	one bypass/disconnect (47)	Crime so (see	ene description list below) (48)	/ (49)	
Crime Lab report Da (50)	ste (51) (52) (53) (54)	Criminalist name			
Crime Scene D	escription				
	= Fouled premises = Used facilities	5 = Consumed food 6 = Other	/drink (state)		22]

	Card	type $\frac{7}{(1)(2)}$ Primary Fel	ony Offense (3) (4)	R.D. Report No. (5) (6) (7) (8) (9) (10) (11)
	Crin	inal Investigation Sources Checked:	Date Checked (month/day)	Response
	[1]	Vehicle registration (1028)	(12) (13) (14) (15)	Useful lead (16) Vehicle registered Vehicle stolen (17) Vehicle stolen (18)
	[2]	Warrant information	(19) (20) (21) (22)	Useful lead (23)
	[3]	Crime file run-person	(24) (25) (26) (27)	Useful lead Suspect(s) ID (29)
	[4]	Crime file - vehicle	(30) (31) (32) (33)	Useful load Vehicle stolen Vehicle linked (34) (35) to suspect (36)
	[5]	Field contact report	(37) (38) (39) (40)	Useful lead (41)
날	[6]	Crime re=cap logs	(42) (43) (44) (45)	Useful lead (46)
	[7]	Drivers license physical data	(17) (48) (49) (50)	Useful lead(51)
	[9]	TT to DOJ/stolen property	(52) (53) (54) (55)	[8] Stolen article Positive link recovery run (56)
	Ç [10]	Pin map data	(58) (59) (60) (61)	Useful feedback (62)
	[11]	Firearms query	(63) (64) (65) (66)	Stolen [14] Registered to suspect (67)
	[12]	CORPUS information	(69) (70) (71) (72)	Positive response (73)
	[15]	Other(state)	(74) (75) (76) (77)	Useful load [] Case code
	ęļ.	1		*

* If date not known, use XXXX,



Card type 8 Primary Felony Offense (3) (4)

R.D. Report No.

(12)

General Offender Description: (see list # 1)

Age group (13)

(14)

(15)

Number of Offenders (16)

Duration of time subject in contact or view by reporting party (in minutes) (17)(18)(19)

(Indicate less than One minute by *)

Detailed Physical Description Provided on Subject: Height Weight Eyes Hair Scars

(20)

(21)

(22) (23)

moles, warts (25)

Tattoos (25)

Hair style/ (27)color (28)

Earg (29)

Glasses (30)

characteristics (31)

(24)

Teeth (34)

Nose (35)(36)

Facial hair (37)

(Y/N)

Hands (38)

Torso Arms (39)

Legs (40)

Build (41) (overall) (42)

S = Slight M = Medium H = Heavy

U = Unknown

Detailed Description of Subject's Clothing Provided *: Headgear

Top coat/ __ rainwear (44) Jacket

Shirt/ (45)blouse (46)

Pants

(48)

Skirt/ dress (49)

Footvear (50) Uniform

(51)

Race

Suit

(43)

(52)

Head/face covering (53) Gloves

(54)

Other

(55)

Color

(47)

(56)

(state=) .

List # 1 -- General Offender Description

Age group

W = White M = Mexican

C ≡ Chinese

N = Negro I = Amer. Indian

0 = Other (state)

À = Adult J ≥ Juvenile

F = Female

225

224

† Override by * if there is a physical anomaly indicated such as amputation, or unusual deformities.

A Override each article of clothing with * for color given.

WITNESS/VICTIM CONFRONTATION

Description of Subject's Speech:

Words spoken

Silent/note

Notable accent

Speech defect

(57)

passed

(58)

(59)

Description of Subject's Actions that Seemed Unusual:

Psychological state

Pretended to be

(see list # 1) (61)

(see list # 2) (62)

Offender's Associates/Movements:

Suspect associates

Places suspect(s) ___

Direction of flight

named/indicated (63)

frequented named $(\overline{64})$

provided

Offender movement by (sec list # 3) (66)

Suspects known to ___ (see list # 4) (67)

Suspects previously seen by

Suspect(s) named

(see list # 4)

(68)

(see list # 5) (69)

List # 1 Psychological State	List # 2 Pretended	to be	List # 3 Offender Movement
<pre>1 = Calm 2 = Nervous 3 = Under influence of drugs/ intoxicated 4 = Violent</pre>	<pre>1 = Ask direction 2 = Ailing 3 = Customer 4 = Repair/delivery 5 = Seeking someone</pre>	6 = Panhandling 7 = Salesman 8 = Asked for something 9 = Other	<pre>1 = Foot 2 = Auto 3 = Bicycle 4 = Unknown 5 = Other</pre>
List # 4 Suspect Known/Seen By 1 = Victim(s) 2 = Witnesses 3 = Citizen informant 4 = Police informant 5 = Police surmise 6 = Other (state)	1 = Real na	own as (AKA)	

suspect identified)

Card type $\frac{9}{(1)}$ $\frac{9}{(2)}$ Primary Felony Offense ____ R.D. Report No. (5) (6) (7) (8) (9) (10) (11) (3)(4)(12)Name of Suspect (last name first) Identification from: Crime file description ____ Photos (14) (13)Independent witness/victim ID Other in vicinity (16) (17) (18)(state) Arrest made on basis of F.C. Arrest Criteria: Vehicle check Possession of stolen (primary) (suspicious person) (19) (20)property (21)Pickup arrest based upon: a) Radio broadcast ___ b) Daily patrol briefing Suspect named (22)(23)information Cal. license no. of involved vehicle Stolen ___ Linked to suspect (25)(26)(27)(28)(29)(30)(31)(32)Previous citations Linked when in custody Linked via Pawn Linked from other $\overline{(33)}$ for other offense (34)transaction (35) agency tips Linked by Associates __ Polygraph Witness/victim later locates suspect (P/F) $\overline{(38)}$ 229(39)**228**

TENTH CARD -- SUSPECT TRIMINAL HISTORY (Note: Attach form for each of multiple suspects charged)

. ;	Card type $\frac{1}{(1)} \frac{0}{(2)}$ Primary	Felony Offense (3) (4)	R.D. Report No	(5) (6) (7) (8)	(9) (10) (11)	Offender Code (12	2)
#	Name(last name first)	P.F	.N. (13)(14)(15)(1 6)(17)(18)(19)(2() (21)(22)		
	CII No. (23)(24)(25)(26)(27)(28)(29	(30) CEN or Juvenile No.(31)(32) (33) (34) (35) (36) (37) (38) (39) (40		tatus at time rest (list #1)(4	<u>ī</u>)
	D.O.B. (42)(43)(44)(45)(46)(47)	Date of arrest (48)(49)(50)(Time of 51)	arrest (52)(53)(54	0n (Y,	acene /N) (58)	
:	Location of arrest: Oakland' (57)	Other(58) (sta	te)		e e a era sumanna (Alexandra e e e e e	consumment of Method to As	er c 1,255 M
175	Age at time of arrest (59)(60)	Race as determined from an (sec list #2)	rrest and booking	g record			
		ect address			(Census)	1
• Title sale sale i	Oakland Other Alameda (62) County (63)	Contra Costa San F (64)	rancisco(65)	Calif, <u>(66)</u>	Other (67)	(state)	<u>- \-</u>
	First offense charged (98)(69)	Age at time of firs	t arrest (70)(71)				
	List 1 CJS Status 1 = On Parole 2 = On Probation 3 = Out on bail 4 = On own recog. 5 = Escapee 6 = No priors	List 2 Race W = White N = Negro M = Mexican I = Amer. Indian C = Chinese J = Japanese Ø = Other (s	tate)				,



Carre		1	1	
Card	type	<u></u>	<u>_</u>	
		(1)	(2)	

Primary Felony Offense ____ $\overline{(3)}$ $\overline{(4)}$

R.D. Report No. (5) (6) (7) (8) (9) (10) (11)

Offender Code (12)

Prior Offenses:

(see list below)

(13)(14)

Date (15)(16)(17)(18)(19)(20) (month day

year)

Location (21)

Disposition ___

R.D. No.

(30)(31)

Date (32)(33)(34)(35)(36)(37) Location (38) Disposition

(47)(48)

(49)(50)(51)(52)(53)(54)

Location (55) Disposition ___

R.D. No.

(57)(58)(59)(60)(61)(62)(63)

(64)(65)

Date

(66)(67)(68)(69)(70)(71)

Location (72) Disposition ___

(73)

R.D. No.

(74)(75)(76)(77)(78)(79)(80)

Type Prior Offenses]	'ype	Prior	Offer	ises
---------------------	---	------	-------	-------	------

10 = Pursesnatch

11 = Shoplifting

l = Oakland

Location

2 = Other Alameda County

Disposition

1 = Conviction/incarcerated 2 = Probation

3 = Parole

3 = Felony assault 12 = Theft-Other 4 = Burglary (state)

3 = Contra Costa County

(City)

4 = Juvenile disposition

5 = Auto theft

1 = Strongarm robbery

2 = Armed robbery

13 = Narcotics and drugs

4 = San Francisco

5 = Other

6 = Homicide, willful

14 = Stolen property

5 = California

7 = Forcible rape 8 = Attempted rape 15 = Vehicle laws 16 = Othe<u>r</u> 6 = Other

(state)

9 = Theft from person

(state)

Card type $\frac{1}{(1)} \frac{2}{(2)}$

Primary Felony Offense $\frac{}{(3)}$ $\frac{}{(4)}$

R.D. Report No.

(5) (6) (7) (8) (9) (10) (11)

Appendix D DATA COLLECTION AND PROCESSING METHODOLOGY

APPENDIX D. DATA COLLECTION AND PROCESSING METHODOLOGY

1. Data Collection Procedures

- a. <u>Felony crimes coded</u>. We have already stated that the felony crimes selected for development of case selection rules were: robbery, rape, ADW, and car theft. The reasons for selecting these crime categories were:
 - They are all Part I crimes as classified by the FBI
 Uniform Crime Reports.
 - The only Part I crimes not represented are burglary and homicide. Burglary had been examined in a previous felony investigation research project.* Homicide was not selected because, regardless of the information available, police departments investigate all homicides owing to the seriousness of the offense.
 - ADW was selected from the category of felonious assault because it is the highest-volume assault category in Oakland.

We coded data covering a three-month period: July, August, and September 1974. The cases were classified into the following three categories:

• <u>Cleared</u>. These cases were those for which the OPD took one of the following formal clearances [see Appendix B Figure B-9(a)].

k B. Greenberg et al., op. cit.





- Arrest and prosecution (includes cases where warrants had been issued).
- Prosecuted for another offense.
- D.A. citation issued.
- Prosecuted by outside department.
- Turned over to Juvenile Authority (juvenile disposition).
- Reprimanded and released (juvenile disposition).
- Notice to appear (juvenile disposition).
- Cleared-other. These cases included some of those for which the OPD took either a "Complainant Refuses To Prosecute" or "Complaint Refused by District Attorney" clearance. We placed only the cases where a suspect was named into the cleared-other category. The OPD also uses the "Complainant Refuses To Prosecute" clearance in some cases where the complainant does not cooperate-for example, by not returning the investigator's phone calls or refusing to come to the OPD to view mugs of possible suspects. Such cases were placed into the uncleared category.
- <u>Uncleared</u>. Included in the uncleared category were cases fulfilling one of the following criteria:
 - Cases classified as "Complainant Refuses To Prosecute," where a suspect was not named.
 - Cases where the investigator filed the case without a clearance. Typically, the investigator stated he was "filing the case pending further investigative leads."
 - Cases where there was no evidence of investigative attention.





- b. <u>Sampling procedure</u>. To identify the report numbers of the cases to be coded, we transcribed the City of Oakland OPD tapes for the time period into a suitable format that grouped report numbers by felony categor thus facilitating the drawing of specific reports. Because of the large number of crimes reported during the time period, it was not necessary to code all the ADWs, robberies, and car thefts to achieve an adequate sample size. We disregarded cases classified as "unfounded" by the OPD as not being germane to the project objectives. Our general sampling criteria were to code all cleared cases and a random sample of uncleared cases, as follows:
 - Strong-arm and armed robbery. All cleared and clearedother cases and approximately one-fourth of all uncleared cases were coded.
 - Purse snatch and theft from person. All cleared and cleared-other and approximately one-third of all uncleared cases were coded.
 - Rape. All rape cases were coded.
 - ADW. Of the cases in the cleared category, two-thirds of the "Arrest and Prosecution" and all the other cases were coded. One-third of the cleared-other cases were coded. All the uncleared cases were coded. (This sampling procedure differs from those for the other felony categories because the greater number of cases fall into cleared and cleared-other rather than into uncleared categories, as is the case for the other crimes under consideration.)
 - Car theft. All the cleared and cleared-other cases were coded, together with one-sixth of the uncleared cases.



All the tables presented in this report reflect weighting factors that were applied to the sample drawn so that they would reflect the total number of felony cases reported during the three-month period.

- c. <u>Data coding form</u>. An 11-page data coding form was developed to record the information from the various OPD reports (see Appendix C for the form used). The data collection form provided for coding of information in the following areas:
 - General information. Felony offense, time of occurrence, case disposition, beat, and census.
 - Personnel involved. Reporting, investigating, and arresting officers, evidence technician, and dates of involvement.
 - Crime scene. Location of crime and type of facility involved.
 - General incident descriptors. Who reported
 offense; age, sex and race of principal report ing individuals; condition of victim; and
 weather conditions.
 - · Property taken.
 - Weapon used.
 - Vehicle used. Whether description, color, and/or license number were provided.
 - · Physical evidence present.
 - Investigative resources utilized. Whether
 various computer or manual information systems were utilized and whether they provided
 information useful in the investigation.





- Offender descriptors. Information elements on multiple offenders to a maximum of five, such as: age, sex, and race; duration of time offender was in contact with or in view by reporting party; physical description and clothing description provided on offender; information regarding offender's associates and movements (e.g., whether offender was known to victim, offender's name was given, or offender's direction of flight was provided).
- Means by which the offender was identified and how arrest was effected.
- Suspect criminal history. Date of birth, date and time of arrest, residence at time of arrest, age at time of first arrest, and so forth.
- Prior offenses. Type, date, location, and disposition of the suspect's prior offenses.

The data collection form was designed, pretested, and modified to reflect the type of information available in the OPD files.

d. <u>Data coding procedures</u>. After the felony case report numbers had been identified, the cases were pulled from OPD files for coding. Whenever possible, we used the CID files because they were generally the most complete. When the case files could not be located in the CID, we consulted the Records Division files. In all cases, the entire file was read: the initial offense report, the arrest report, the follow-up investigation report, supplemental statements, evidence technician reports, the crime analysis, EDP printouts, and the like.

Frequently, information on prior criminal involvement of identified suspects was not contained in the case files. In these instances, we used several other OPD resources to obtain the information: criminal history diazo microfiche files located in the CAS, juvenile records located in the Youth Services Division, and the Alameda County CORPUS information. All subject identities were suppressed in the data processing procedures.

2. Data Processing Procedures

Essentially the same data processing procedures were followed for each of the four felony categories. These procedures were designed with several goals in mind:

- Successive reduction in the number of variables under consideration. As discussed in the preceding section, the data coding form provided for close to 1000 variables.
 A primary goal of the data processing was to provide a means for reducing the number of variables considered for input to a decision model construct.
- Understanding the differences between cases that were cleared and those that remained uncleared. All our variables were considered in light of their association with clearance. Our goal, therefore, was not merely to be able to describe the general characteristics of the four felony types but rather to be able to state what distinguished the cleared from the uncleared cases.
- Forming ways of predicting whether a case will be cleared or remain uncleared. We were interested in developing models which when applied to cases of different felony types would be able to predict with a high degree of accuracy whether a case would be cleared.

Our data processing procedures were statistically based, rather than anecdotal in nature. In other words, we were looking at generalized investigations according to felony types rather than at individual cases. Our objective was to develop generalized models that would predict whether a case taken at random would be cleared rather than in investigating in depth the factors that led to the solution of a particular case. Clearly, at times, certain factors that are not significant in a statistical sense lead to case closure. We, however, are more concerned with factors that can predict case clearance with a high degree of accuracy in a large sample of cases. This view is consistent with police management practices which must be applied to the high volume of reported crimes.

The following tasks were undertaken in the data processing procedures for each of the four felony types:

- Keypunching and cleaning up data. The obvious first step in the data processing was to keypunch the data and eliminate coding and keypunch errors. Also, at this stage we were able to make the first reduction in the size of the data base by determining which variables never or rarely appeared.
- Setting up SPSS files. The Statistical Package for the Social Sciences (SPSS) was chosen as the primary medium for the analysis of the felony data. A major reason for this choice is the flexibility of the data management facilities available in this package. The data in an SPSS file can easily be recoded and combined, as well as written out in a variety of forms for use in other analyses.
- Running cross tabulations. Extensive cross tabulations
 were run using an SPSS subprogram and an SRI-developed
 program. These cross tabulations were carefully analyzed



to determine the variables that appeared to be associated with the cleared or uncleared cases. For example, the variables suspect named and suspect known were obviously associated with clearance. In other cases, a more subtle association appeared.

- Deriving Pearson correlation coefficients. Another measure of the relation between two variables is the correlation coefficient. Correlation coefficients were calculated for over 100 variables for each of the felony types in order to determine the statistical importance of the relationships observed in the cross tabulations. The SPSS subprogram used gives the following data for each correlation coefficient calculated:
 - The actual correlation coefficient. This is a number that varies between -1 and 1. We set up the data so that a number close to 1 would indicate a high positive correlation with clearance. A number close to -1 indicates a high negative correlation with clearance. A number close to 0 indicates that the variable had little correlation with clearance.
 - The number of cases used in the calculation, depending on the number of missing values for the variable pair.
 - The level of statistical significance of the coefficient. The closer this number is to 0, the higher the degree of statistical significance.
 - Table D-1 illustrates the variables chosen initially for robbery analysis and the correlation these variables show with arrest for armed and strong-arm robbery (Table D-2).



Table D-1

INITIAL LIS

7 ROBBERY VARIABLES

- VAROO4 Evidence technician at crime scene
- VAR005 Crime location-street
- VAR006 Crime location-building
- VAR007 Crime reported by witness
- VAROO8 One reporting individual
- VAROO9 Two reporting individuals
- VAR010 Three or more reporting individuals
- VAR011 Adult victim
- VAR012 Juvenile victim
- VAR013 Female victim
- VAR014 Male victim
- VARO15 White victim
- VAR016 Black victim
- VARO17 Victim of other race
- VAR018 Victim lucid
- VAR019 Victim cooperative
- VAR020 Handgun used
- VAR021 Knife used
- VAR022 Other weapon used
- VARO23 Sexual aberrations indicated
- VAR024 Vehicle used
- VARO25 Description of vehicle given
- --- VARO26 Color of vehicle given
 - VAR027 License number of vehicle given
 - VAR028 Cash, negotiables, taken
 - VAR029 Credit cards taken
 - VAR030 Less than \$100 taken
 - VAR031 \$100-200 taken
 - VAR032 \$200-500 taken
 - VAR033 \$500-1000 taken
 - VAR034 \$1000-2000 taken
 - VAR035 More than \$2000 taken
 - VARO36 Victim invited offender in
 - VARO37 Attack against property
 - VARO38 Attack against person
 - VAR039 Fingerprints taken
 - VARO40 Fingerprints match
 - VARO41 Weapons as evidence
 - VARO42 Weapons match
 - VARO43 Clothing as evidence

Table D-1 (Continued)

```
VARO44 - Clothing match
VARO45 - Other physical evidence
VAR046 - Other physical evidence match
VARO47 - Vehicle registration check made
VARO48 - Vehicle registration check--useful lead
VAR049 - Vehicle registered to suspect
VARO50 - Vehicle stolen
VARO51 - Crime file run-person
VARO52 - Crime file run-person--useful lead
VAR053 - Crime file run-vehicle
VAR054 - Field contact report
VAR055 - Adult offender
VARO56 - Juvenile offender
VAR057 - Female offender
VARO58 - Male offender
VARO59 - White offender
VARO60 - Black offender
VARO61 - Mexican-American offender
VAR062 - One offender
VAR063 - Two offenders
VAR064 - Three or more offenders
VARO65 - Less than 1 minute contact between victim and offender
VARO66 - 1-10 minutes contact between victim and offender
VARO67 - 11-30 minutes contact between victim and offender
VARO68 - Greater than 30 minutes contact between victim and offender
VAR069 - Height of offender given
VAR070 - Weight of offender given
VARO71 - Eyes of offender described
VAR072 - Hair of offender described
VAR073 - Offender described as wearing glasses
VAR074 - Teeth of offender described
VAR075 - Sum of physical descriptors given (not a binary variable)
VARO76 - One or two physical descriptors given
VAR077 - Three physical descriptors given
VARO78 - Four or more physical descriptors given
VAR079 - Offender described as wearing jacket
VARO80 - Offender described as wearing shirt/blouse
VARO81 - Offender described as wearing pants
VARO82 - Sum of clothing descriptors given (not a binary variable)
```



Table D-1 (Concluded)

```
VARO83 - One or two clothing descriptors given
VARO84 - Three clothing descriptors given
VAR085 - Four or more clothing descriptors given
VAR086 - Words spoken by offender
VAR087 - Offender silent/note passed
VARO88 - Offender described as violent
VARO89 - Offender pretended to be: asking directions, ailing,
         customer, repair/delivery, seeking someone, panhandling,
         salesman, asked for something, other
VAR090 - Suspect's associates named/indicated
VAR091 - Places suspect frequented named
VARO92 - Direction of flight provided
VAR093 - Offender movement by automobile
VAR094 - Offender movement by foot
VAR095 - Suspect known to: victim(s), witnesses, citizen informant,
         police informant, police surmise, other
 VARO96 - Suspect previously seen by: victim(s), witnesses,
          citizen informant, police informant, police surmise,
          other
 VAR097 - Suspect named: real name, also known as, partial, nickname
 VAR098 - Less than 1 hour between occurrence and report of crime
 VAR099 - One to 2 hours between occurrence and report of crime
 VAR100 - More than 2 hours between occurrence and report of crime
 VAR101 - Crime occurred between 0001 and 0400 hours
 VAR102 - Crime occurred between 0401 and 0800 hours
 VAR103 - Crime occurred between 0801 and 1200 hours
 VAR104 - Crime occurred between 1201 and 1600 hours
 VAR105 - Crime occurred between 1601 and 2000 hours
 VAR106 - Crime occurred between 2001 and 2400 hours
 VAR107 - White offender and white victim
 VAR108 - White offender and black victim
 VAR109 - Black offender and black victim
 VAR110 - Black offender and white victim
 VAR111 - Offender and victim same race
```



Table D-2
PEARSON CORRELATION COEFFICIENTS: STRONG-ARM/ARMED ROBBERY VERSUS ARREST

Vari	able Pair	Vari	able Pair	<u>Vari</u>	able Pair	Vari	able Pair	Var L	able Pair	Vari	able Pair
Arrest	0.1749	Arrest	0.0070	Arrest	-0.0149	Arrest	0.1405	Arrest	-0.1605	Arreat	0.0200
with	N(605)	with	N(588)	with	N(588)	with	N(605)	with	N(595)	with	N(595)
VAROO4	Sig 0.000	VAROOS	Sig 0.865	VAROO6	Sig 0.719	VAROO7	Sig 0.001	VAROOB	Sig 0.000	VAROO9	Sig 0.627
Arrest	0.2604	Arrest	-0.0602	Arrest	0.0602	Arrest	-0.0160	Arrest	0.0160	Arrest	-0.1003
with	N(595)	with	N(603)	with	N(603)	with	N(603)	with	N(603)	with	N(596)
VARO10	Sig 0.000	VARO11	Sig 0.139	VARO12	Sig 0.139	VARO13	Sig 0.695	VARO14	Sig 0.695	VARO15	Sig 0.014
Arrest	0.0766	Arrest	0.0495	Arrest	0.0918	Arrest	0.1983	Arrest	-0.1131	Arrest	0.0372
with	N(596)	with	N(596)	with	N(379)	with	N(346)	with	N(605)	with	N(605)
VARO16	Sig 0.062	VARO17	Sig 0.228	VARO18	Sig 0.074	VARO19	Sig 0.000	VARO20	Sig 0.005	VARO21	Sig 0.361
Arrest	0.1069	Arrest	0.1625	Arrest	0.1387	Arrest	0.1603	Arrest	0.1414	Arrest	0.3190
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO22	Sig 0.009	VARO23	Sig 0.000	VARO24	Sig 0.001	VARO25	Sig 0.000	VARO26	Sig 0.000	VARO27	Sig 0.000
Arrest	-0.0971	Arrest	-0.0650	Arrest	-0.0242	Arrest	-0.0123	Arrest	0.0355	Arrest	0.0106
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO28	Sig 0.017	VARO29	Sig 0.110	VARO30	Sig 0.553	VARO31	Sig 0.762	VARO32	Sig 0.384	VARO33	Sig 0.794
Arrest	0.0135	Arrest	-0.0445	Arrest	0.1178	Arrest	-0.2260	Arrest	-0.0053	Arrest	0.1031
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO34	Sig 0.740	VARO35	Sig 0.274	VARO36	Sig 0.004	VARO37	Sig 0.000	VARO38	Sig 0.897	VARO39	Sig 0.011
Arrest	0.1090	Arrest	0.1132	Arrest	0.2898	Arrest	0.1425	Arrest	0.2681	Arrest	0.2104
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO40	Sig 0.007	VARO41	Sig 0.005	VARO42	Sig 0.000	VARO4)	Sig 0.000	VARO44	Sig 0.000	VARO45	Sig 0.000
Arrest	0.3077	Arrest	0.1663	Arrest	0.2445	Arrest	0.1283	Arrest	0.0238	Arrest	-0.1842
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO46	Sig 0.000	VARO47	Sig 0.000	VARO48	Sig 0.000	VARO49	Sig 0.002	VARO50	Sig 0.559	VARO51	Sig 0.000
Arrest	0.1543	Arrest	-0.0617	Arrest	0.2260	Arrest	-0.1254	Arrest	0.1254	Arrest	0.0799
with	N(605)	with	N(605)	with	N(605)	with	N(574)	with	N(574)	with	N(604)
VARO52	Sig 0.000	VARO53	Sig 0.129	VARO54	Sig 0.000	VARO55	Sig 0.003	VARO56	Sig 0.003	VARO57	Sig 0.050

Variable Pair		Variable Pair		Variable Pair		Variable Pair		Variable Pair		Variable Pair	
Arrest	-0.0799	Arrest	0.1309	Arrest	-0.0957	Arrest	-0.0108	Arrest	-0.0210	Arrest	0.0258
with	N(604)	with	N(604)	with	N(604)	with	N(604)	with	N(605)	with	N(605)
VARO58	Sig 0.050	VARO59	Sig 0.001	VARO60	Sig 0.019	VARO61	Sig 0.791	VARO62	Sig 0.607	VARO63	Sig 0.527
Arrest	-0.0059	Arrest	-0.0720	Arrest	-0.0839	Arrest	0.1234	Arrest	0.1845	Arrest	-0.1472
with	N(605)	with	N(560)	with	N(560)	with	N(560)	with	N(560)	with	N(605)
VARO64	Sig 0.884	VARO65	Sig 0.089	VARO66	Sig 0.047	VARO67	Sig 0.003	VARO68	Sig 0.000	VARO69	Sig 0.000
Arrest	-0.0060	Arrest with VARO71	0.0857	Arrest	-0.0431	Arrest	0.0507	Arrest	-0.0121	Arrest	-0.0881
with	N(605)		N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VARO70	Sig 0.884		Sig 0.035	VARO72	Sig 0.290	VARO73	Sig 0.213	VARO74	Sig 0.766	VARO75	Sig 0.030
Arrest	0.0179	Arrest	-0.0256	Arrest	0.0057	Arrest	-0.0401	Arrest	0.0258	Arrest	-0.1107
With	N(517)	with	N(517)	with	N(517)	with	N(605)	with	N(605)	with	N(605)
VARO76	Sig 0.684	VARO77	Sig 0.561	VARO78	Sig 0.897	VARO79	Sig 0.325	VARO80	Sig 0.527	VARO81	Sig 0.006
Arrest	-0.1010	Arrest	-0.0427	Arrest	=0.0002	Arrest	-0.0701	Arrest	0.0523	Arrest	0,0014
with	N(605)	with	N(468)	with	N(468)	with	N(468)	with	N(605)	with	N(605)
VARO82	Sig 0.013	VARO83	Sig 0.356	VAR084	Sig 0.996	VARO85	Sig 0.130	VARO86	Sig 0.199	VARO87	Sig 0.973
Arrest	0.1260	Arrest	0.0107	Arrest	0.0634	Arrest	0.1602	Arrest	0.0571	Arrest	0.1126
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(413)
VARO88	Sig 0.002	VARO89	Sig 0.793	VARO90	Sig 0.120	VARO91	Sig 0.000	VARO92	Sig 0.160	VARO93	Sig 0.022
Arrest.	-0.1666	Arrest	0.2487	Arrest	0.3087	Arrest	0.1972	Arrest	-0.0302	Arrest	=0,0678
with	N(413)	with	N(605)	with	N(605)	with	N(605)	with	N(588)	with	N(588)
VARO94	Sig 0.001	VARO95	Sig 0,000	VARO96	Sig 0.000	VARO97	Sig 0.000	VARO98	Sig 0.465	VARO99	Sig 0.100
Arrest	0.0762	Arrest	-0.0192	Arrest	-0.0647	Arrest	0.1881	Arrest	-0.0111	Arrest	-0.0248
with	N(588)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VAR100	Sig 0.065	VAR101	Sig 0.638	VAR102	Sig 0.112	VAR103	Sig 0.000	VAR104	Sig 0.785	VAR105	Sig 0.543
Arrest	-0.0179	Arrest	0.1572	Arrest	0.1543	Arrest	0.0333	Arrest	-0.1206	Arrest	0.0730
with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)	with	N(605)
VAR106	Sig 0.660	VAR107	Sig 0.000	VAR108	Sig 0.000	VAR109	Sig 0.413	VAR110	Sig 0.003	VARIII	Sig 0.073

A value of 99.0000 is printed if a coefficient cannot be computed.



• Conducting discriminant analyses. Variables showing a degree of correlation with clearance and a reasonable level of statistical significance were then selected for discriminant analysis. Because discriminant analysis assumes that the variables are independent from one another, extensive recoding was done to establish this independence. For example, instead of four variables—vehicle used or taken, description given, color given, and license number given—one vehicle variable was created with values on a sliding scale. (If a vehicle was used or taken, a score of 1 was assigned; if the vehicle was described, a score of 2 was assigned; if the color was given, a score of 3 was assigned; and if a license number was given, a score of 4 was assigned.)

Discriminant analysis was the technique chosen for the final development of the model, because it is particularly well suited for separation of groups be - ! on the relative importance of the variables. The disc. calculation forms a linear combination of the discriminating variables called the discriminant function. The weighting coefficients used in this function are a measure of the relative value of the variable in separating the groups. With this knowledge a classification coefficient (weight factor) could be derived for use in the case follow-up decision model. We chose to use a combination of the BMD and SPSS packages for the analysis. BMD provides output that is formatted in such a way that the calculated values of the discriminant coefficients are more easily traced to the actual values of the variables. SPSS, owing to its superior data management capabilities, facilitated extensive experimentation in



variable design and recoding. The values calculated by the two packages are not significantly different. (See Appendix E for a technical discussion of discriminant analysis.)

Analyzing offender data. The criminal history data processing involved the use of several SPSS and SRI-developed computer programs. A variety of other statistical summary and analysis techniques were used to extract inferences and conclusions from the data.



Appendix E

DISCUSSION OF DISCRIMINANT ANALYSIS METHODOLOGY

APPENDIX E. DISCUSSION OF DISCRIMINANT ANALYSIS METHODOLOGY

1. General

The objective of discriminant analysis is to provide a statistical basis to distinguish between members of two or more groups (or populations) with acceptable probabilities of being correct. The technique involves the sampling of cases with confirmed group membership by obtaining measurements on variables that would characterize the group identity of the cases. These variables are called discriminating variables. On the basis of the value of these variables, "discriminant functions" are constructed that serve as the basis of a "decision rule" to be used in the classification of cases with unknown memberships.

2. The Construction of Discriminant Functions

Let:

g = total number of groups

i = group index

 n_i = sample size of Group i

j = sample index

m = total number of discriminating variables

k = discriminating variable index

 x_{ijk} = value of variable k for the j th sample in Group i.

The maximum allowable number of discriminant functions = min. g - 1, m.

For this project, the groups that we wish to be differentiated for each offense in question are the cleared cases and the uncleared cases. Therefore, only one discriminant function is allowed, which takes the form:

$$Z = \lambda_1 x_1 + \lambda_2 x_2 + \dots \lambda_m x_m ,$$

where Z is the composite score of the discriminant function; $\lambda_1, \lambda_2, \ldots, \lambda_m$ are the weighting coefficients; and x_1, x_2, \ldots, x_m are the values of the m discriminating variables used in the discriminant analysis. These variables may be normalized if desirable.

The vector λ (i.e., λ_1 , λ_2 , ..., λ_m) is derived by solving simultaneously the following set of equations (in matrix notation):

$$A\lambda = d$$
.

where $A = S^1 + S^2$,

$$s^{i} = \begin{bmatrix} s^{i}_{uv} \end{bmatrix}$$
 for $u = 1, 2, ..., m$
 $v = 1, 2, ..., m$

$$S_{uv}^{i} = \sum_{j=1}^{n_{i}} (x_{iju} - x_{i.u}) (x_{ijv} - x_{i.v})$$

and d is a vector of the differences between the means of two groups on the m measurements:

$$d = \left[x_{1,} - x_{2,}\right] = \left[x_{1,1} - x_{2,1}, x_{1,2} - x_{2,2}, \dots, x_{1,m} - x_{2,m}\right]$$

The vector λ may be solved as follows:

$$\lambda = A^{-1}d$$

The resulting vector is a set of weighting coefficients characterizing the most discriminating linear combination of the variables measured.



Depending on the discriminating variables used, the Z score can be computed for:

• The mean values for the discriminant functions for Groups 1 and 2:

$$\overline{Z}_1 = \lambda_1 x_{1.1} + \lambda_2 x_{1.2} + \dots + \lambda_m x_{1.m}$$
 $\overline{Z}_2 = \lambda_1 x_{2.1} + \lambda_2 x_{2.2} + \dots + \lambda_m x_{2.m}$

For each mean-value discriminant function, the variance $V(Z_i)$ and standard deviation $\sqrt{V(Z_i)}$ can also be computed.

• Each sample in Groups 1 and 2:

3. Tests

Two tests are commonly used for estimating the statistical significance of an analysis.*

Mahalanobis' D²:

$$D^2 = (n_1 + n_2 - 2) \sum_{i=1}^{m} \sum_{j=1}^{m} a^{ij} (x_{1.i} - x_{2.i}) (x_{1.j} - x_{2.j})$$

where $[a^{ij}]$ are elements of A^{-1} .

This test is to estimate the squared distance between the mean values of the discriminating variables of the two groups. It is desirable that ${\tt D}^2$ be maximized.

^{*}See Dixon, p. 216. A selected bibliography is given at the end of this appendix.





The F-test:

$$F(m_1m_1 + n_2 - 1 - m) = \frac{n_1n_2(n_1 + n_2 - m - 1)}{m(n_1 + n_2)(n_1 + n_2 - 2)} \cdot D^2$$

This test requires the assumption that the sample measurements have a multivariate normal distribution. It is used for testing the similarity of the variances between the two populations.

4. The Selection of a Cutting Point

Under a two-group classification scheme, we can view the system as having two distributions along the same axis, with an overlapping area (see Figure E-1). Point C is a cutting point, which is an arbitrary separation point between the two groups:

$$j \in \begin{cases} G_1 & \text{if } Z_j < C \\ G_2 & \text{if } Z_j > C \\ G_1 \text{ or } G_2 & \text{if } Z_j = C \end{cases}$$

where \mathbf{G}_1 and \mathbf{G}_2 denote Group 1 and Group 2, respectively.

The area E_1 represents the probability of case misclassification for Group 1, i.e., of the cases being classified as belonging to Group 2 when, in fact, they belong to Group 1. Similarly, the area E_2 represents the probability of misclassification of cases having membership in Group 2.

The position of C is determined by the risk values that the analyst places on the consequences of misclassification. Assuming normal distribution, the area $\rm E_1$ or $\rm E_2$ may be computed by transforming C into a unit-normal deviate $\rm K_c$.



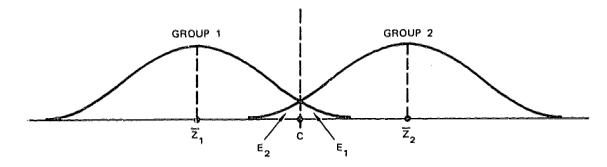


FIGURE E-1 DISTRIBUTIONS OF Z-SCORES IN TWO GROUPS

For E₁:

$$K_{c} = \frac{C - \overline{Z}_{1}}{\sqrt{V(Z_{1})}}$$

For E₂:

$$K_{c} = \frac{C - Z_{2}}{\sqrt{V(Z_{2})}}$$

If the distributions of the two populations are approximately equal, and the loss function of misclassification between the two groups is identical, then a cutting point placed midway between \overline{Z}_1 and \overline{Z}_2 , i.e., C=1/2 $(\overline{Z}_1+\overline{Z}_2)$ (hence $E_1=E_2$), would be desirable. Otherwise, the frequency distribution of the known cases under each group, as well as the population size of each group, should be analyzed before the selection of a cutting point.

5. The Determination of a Decision Rule

In the foregoing discussion, all the measurements for the discriminating variables are taken from samples with known dispositions (cleared or uncleared). The determination of a cutting point is then primarily

based on historical data. However, the utility of the discriminant analysis lies not only in establishing historical relationships but in predicting the identity of cases in which the group memberships are not known. The same discriminating variables used in constructing the discriminant function will be used to measure the new cases and to calculate the Z scores, using the weighting coefficients developed from historical data. The Z score for each new case is then compared with Cutting Point C in order to establish the probable group identity of the new case.

Since Point C is relative to the scale chosen for the discriminant function, multiplication of the entire discriminant function by a scalar would automatically change the magnitude of C but would not alter the relationship between the cutting point and the Z scores. Therefore, it is often operationally convenient to select a modified value of C that is a positive integer, e.g., 10, instead of the original C value, say, -0.0192. Likewise, the weighting coefficients should also be multiplied by the same scalar in the computation for the Z scores; thus the Z scores would be on the same scale with the threshold value C. Let C' and Z_j' denote the transformed cutting point and the transformed Z score for new Case j, respectively, then the decision rule may be shown as follows:

Case j belongs to:
$$\begin{cases} \text{Group 1} & \text{if } C' > Z'_{j} \\ \text{Group 2} & \text{if } C' < Z'_{j} \\ \text{Indifferent if } C' = Z'_{j} \end{cases}$$

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